

Analysis 2 Lecture Notes
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Chapter 1

Maximum and Minimum of Multivariable Functions

For a C^3 function (in a neighborhood of a in \mathbb{R}), by Taylor's Theorem

$$f(a+h) = f(a) + f'(a)h + \frac{1}{2}f''(a)h^2 + \underbrace{\frac{1}{6}f''' \left(\begin{array}{c} \text{some point} \\ \text{between} \\ a \text{ and } a+h \end{array} \right) h^3}_{\text{Remainder term } r(h)} \\ \frac{r(h)}{h^2} \rightarrow 0 \text{ as } h \rightarrow 0$$

Suppose $f'(a) = 0$ “ a is a critical point of f ”. Then

$$\frac{f(a+h) - f(a)}{h^2} = \frac{1}{2}f''(a) + \frac{r(h)}{h^2}$$

If $f''(a) > 0$ then f has a local minimum at a because choose $\delta > 0$ such that $|h| < \delta$, $\left| \frac{r(h)}{h^2} \right| < \frac{1}{2}f''(a)$. Then $RHS > 0 \forall h$ such that $|h| < \delta$ and so for $h \in (-\delta, \delta)$, $f(a+h) > f(a)$ i.e. $f(a)$ is minimum value of f in the neighborhood $(a-\delta, a+\delta)$. Similarly $f''(a) < 0$ then f has a local maximum at a .

We want to find an analogy of this for multivariable case

$f : (\text{open } U \text{ in } \mathbb{R}^n) \rightarrow \mathbb{R}$ a C^3 function. Then for $h \in$ some open neighborhood W of origin, $a+h \in U$

$$\begin{aligned} f(a+h) &= f(a) + f'(a)h + \frac{1}{2}f''(a)(h, h) + \underbrace{\frac{1}{6}f''' \left(\begin{array}{c} \text{some point} \\ \text{between} \\ a \text{ and } a+h \end{array} \right) (h, h, h)}_{\text{Remainder term } r(h)} \\ &\quad \frac{r(h)}{h^2} \rightarrow 0 \text{ as } h \rightarrow 0 \\ &= f(a) + [D_1 \quad \cdots \quad D_n] \begin{bmatrix} h_1 \\ \vdots \\ h_n \end{bmatrix} + \frac{1}{2} \sum_{i,j} D_i D_j f(a) h_i h_j + r(h) \\ &= f(a) + [D_1 \quad \cdots \quad D_n] \begin{bmatrix} h_1 \\ \vdots \\ h_n \end{bmatrix} + \frac{1}{2} [h_1 \quad \cdots \quad h_n] [D_i D_j f(a)] \begin{bmatrix} h_1 \\ \vdots \\ h_n \end{bmatrix} + r(h) \\ &= f(a) + \nabla f(a) \cdot h + \frac{1}{2} h^t \underbrace{[D_i D_j f(a)]}_{\text{Hessian Matrix of } f \text{ at } a} h + r(h) \end{aligned}$$

Definition 1.1: Hessian Matrix of f

Let $f : (\text{open } U \text{ in } \mathbb{R}^n) \rightarrow \mathbb{R}$ such that $\begin{cases} f \text{ is } C^1 \iff \frac{\partial f}{\partial x_i} \text{ are not continuous on } U \\ f'' \text{ exists at } a \end{cases}$ So components of f'' are $D_i D_j f(a)$. Hessian of f at a = Square matrix $[D_i D_j f(a)]$

When f is C^2 , Hessian matrix is Symmetric Matrix

Definition 1.2: Critical Point

Let f be a C^1 function, Open U in $\mathbb{R}^n \xrightarrow{f} \mathbb{R}$. $a \in U$ is called critical point if $f'(a) = 0 \iff \nabla f(a) = 0$

If f has local maximum at a , then along any line through a the same must be hold, so all directional derivative = 0 at a .

Definition 1.3: Non-degenerate Point

If f is C^2 then a critical point a is called non-degenerate if the Hessian, $Hf(a)$ is non-singular i.e. $\det(Hf(a)) \neq 0$

Claim 1.0.1

Symmetric Matrix A is positive (semi)definite $\iff \forall$ nonzero vector $x \in \mathbb{R}^n$, $x^t A x > 0$ (resp. ≥ 0)

Proof. If Part:

$x = \sum_i c_i v_i$. Where v_i is the eigen-basis. Then

$$x^t A x = \left(\sum_i c_i v_i \right)^t A \left(\sum_j c_j v_j \right) = \left(\sum_i c_i v_i \right)^t \left(\sum_j \lambda_j c_j v_j \right) = \sum_i \lambda_i c_i^2 > 0 \quad [v_i^t v_j = \delta_{ij}]$$

Only If Part:

Use $x^t A x > 0$ for $x = v_i$ eigenvector $\implies 0 v_i^t A v_i = v_i^t \lambda_i v_i = \lambda_i$

□

Note:-

Determinant of positive definite matrix > 0 and Determinant of negative definite matrix has sign $(-1)^n$

Theorem 1.1

Let $f : (\text{open } U \text{ in } \mathbb{R}^n) \rightarrow \mathbb{R}$. Suppose f has a local maximum or minimum at a then

- ① If $f'(a)$ exists then $f'(a) = 0$ i.e. a is a critical point.
- ② Suppose in addition to that $f''(a)$ exists then if f has local maximum at a , then $f''(a) \leq 0$ and if f has local minimum at a , then $f''(a) \geq 0$

Proof. ① For $n = 1$ let we have local minimum at a . Then for small $|h|$

$$\left. \begin{aligned} \frac{f(a+h)-f(a)}{h} &\geq 0 & \text{for } h > 0 \\ \frac{f(a+h)-f(a)}{h} &\leq 0 & \text{for } h < 0 \end{aligned} \right\} \text{Thus imply respectively that } f'(a) \text{ must be } \geq 0 \text{ and } \leq 0$$

For $n > 1$ use $n = 1$ in every direction i.e. for function $f|_{a+tv}$ for $t \in$ open interval to conclude $D_v f(a) = 0$ \forall directions. So $f'(a) = 0$

□

② For $n = 1$

$$f''(a) = \lim_{h \rightarrow 0} \frac{f'(a+h) - f'(a)}{h} = \lim_{h \rightarrow 0} \frac{f'(a+h)}{h}$$

Observation: If f has local maximum at a then for $0 < |h| < \delta$, $f(a+h) \leq f(a)$. So by MVT there is k between 0 and h such that

$$\frac{f(a+h) - f(a)}{h} = f'(a+k)$$

Using the observation $f''(a) = \lim_{h \rightarrow 0} \frac{f'(a+k)}{h} \geq 0$

For $n > 1$ applying this to each $f|_{a+tv}$ \forall direction vectors v we get all $D_v^2 f(a) \geq 0$. In terms of Hessian let $v = \sum c_i e_i \implies D_v f = \sum c_i D_i f \implies D^2 f(a) = \sum_{i,j} c_j c_i D_j D_i f(a)$ in a neighborhood of a .

$$D_v^2 f(a) = [c_1 \quad \cdots \quad c_n] H f(a) \begin{bmatrix} c_1 \\ \vdots \\ c_n \end{bmatrix}$$

□

Theorem 1.2

If $f : (\text{open } U \text{ in } \mathbb{R}^n) \rightarrow \mathbb{R}$ is a C^3 function and a is a non-generate critical point of f then

f has a local minimum at a	\iff	H is positive definite
	\iff	All eigenvalues of H are positive
f has a local maximum at a	\iff	H is negative definite
	\iff	All eigenvalues of H are negative
f has saddle-point otherwise	\iff	H is indefinite

Proof. If Part:

We already proved the if direction in [Theorem 1.1](#)

Only If Part:

□