

June 18th

Mid-Term coverage 1 & 2 - exceptions in pdf
Ch2 > Ch1

List of proofs

Recall:

$\sum_{|\alpha| \leq 2}$ ← sum of all indices of length ≤ 2

$$|\alpha|=0$$

$$|\alpha|=|\alpha_1|+|\alpha_2|+|\alpha_3|=0, \alpha=(0,0,0)$$

$$|\alpha|=1, \alpha=(1,0,0) \quad (0,1,0) \quad (0,0,1)$$

$$|\alpha|=2, \alpha=(2,0,0) \quad (0,2,0) \quad (0,0,2)$$

$$(1,1,0) \quad (1,0,1) \quad (0,1,1)$$

§ 2.8 Critical Points

Goal: Find local max/min

1). will prove that if it is a local max/min
 \Rightarrow critical point

2). will find all critical points to be candidates for max/min.
 $\partial_j f(\vec{\alpha}) = 0, \forall j \in \{1, \dots, n\} \leftarrow n \text{ equations}$
 $n \text{ unknowns}$

3). Look at 2nd order derivatives to determine if a max or min.
or higher

Thm: If f has a local max or min at $\vec{\alpha}$, f diff on open set $S \subset \mathbb{R}^n$
then $\nabla f(\vec{\alpha}) = 0$.

Proof: Will show all directional derivatives are 0

Consider $g(t) = f(\vec{\alpha} + t\vec{u})$, \vec{u} as a unit vector.

$$g \text{ has a max at } t=0, g: \mathbb{R} \rightarrow \mathbb{R}$$
$$0 = g'(0) = \frac{d}{dt} f(\vec{\alpha} + t\vec{u}) \Big|_{t=0} = \partial \vec{u} f(\vec{\alpha})$$

$\forall \vec{u}$
all direction derivatives
are 0

$$\Rightarrow \partial_j f(\vec{\alpha}) = 0 \quad \forall j \in \{1, \dots, n\}$$
$$\Rightarrow \nabla f(\vec{\alpha}) = \vec{0}$$

Recall, $f(\vec{a}+h) = f(\vec{a}) + f'(\vec{a})h + \frac{f''(\vec{a})h^2}{2} + R_{\vec{a},2}(h)$
 if $f'(\vec{a})=0$ then depends on small for small h

$$H(\vec{a}) = \begin{pmatrix} \partial_1^2 f(\vec{a}) & \partial_1 \partial_2 f(\vec{a}) & \cdots & \partial_1 \partial_n f(\vec{a}) \\ \partial_2 \partial_1 f(\vec{a}) & \partial_2^2 f(\vec{a}) & \cdots & \partial_2 \partial_n f(\vec{a}) \\ \vdots & \vdots & \ddots & \vdots \\ \partial_n \partial_1 f(\vec{a}) & \partial_n \partial_2 f(\vec{a}) & \cdots & \partial_n^2 f(\vec{a}) \end{pmatrix}$$

↑ Hessian of \vec{a}

$$H_{ij}(\vec{a}) = \partial_i \partial_j f(\vec{a})$$

$$H_{ij}(\vec{a}) = H_{ji}(\vec{a}) \Rightarrow \text{symmetric matrix}$$

$$f(\vec{a}+\vec{k}) = f(\vec{a}) + \sum_{j=1}^n \partial_j f(\vec{a}) k_j + \left[\frac{1}{2} \sum_{i,j=1}^n \partial_i \partial_j f(\vec{a}) k_i k_j + R_{\vec{a},2}(\vec{k}) \right]$$

\downarrow

$$\begin{aligned} &= \frac{1}{2} \sum_{i=1}^n \left(\sum_{j=1}^n H_{ij} k_j \right) k_i \\ &= \frac{1}{2} \sum_{i=1}^n (H \vec{k})_i k_i = \frac{1}{2} \vec{k}^T H \vec{k} \\ &\quad (H_{ij})(k_j) \end{aligned}$$

Putting together, when a CP. $f(\vec{a}+\vec{h}) - f(\vec{a}) = \frac{1}{2} \vec{k}^T H \vec{k} + R_{\vec{a},2}(\vec{k})$
 depends on this term

For all of this f is C^2 & an open set $S \subset \mathbb{R}^n$

$$H = \begin{pmatrix} \partial_1^2 f(\vec{a}) & 0 & \cdots & 0 \\ 0 & \ddots & \ddots & \partial_n^2 f(\vec{a}) \end{pmatrix} = \begin{pmatrix} \lambda_1 & 0 & \cdots & 0 \\ 0 & \ddots & \ddots & \lambda_n \end{pmatrix}$$

$$\lambda_j = \partial_j^2 f(\vec{a})$$

$$\begin{aligned} (k_1, \dots, k_n) \begin{pmatrix} \lambda_1 & 0 & \cdots & 0 \\ 0 & \ddots & \ddots & \lambda_n \end{pmatrix} \begin{pmatrix} k_1 \\ \vdots \\ k_n \end{pmatrix} &= (k_1, \dots, k_n) \begin{pmatrix} \lambda_1 k_1 \\ \vdots \\ \lambda_n k_n \end{pmatrix} = \sum_{j=1}^n \lambda_j k_j^2 \\ &= \vec{k}^T H \vec{k} \\ &= (H \vec{k}) \cdot \vec{k} \end{aligned}$$

$$f(\vec{a} + \vec{k}) - f(\vec{a}) = \frac{1}{2} \sum_{j=1}^n \lambda_j k_j + R_{\vec{a}, 2}(\vec{k})$$

- Sps $\lambda_j > 0, \forall j \in \{1, \dots, n\}$, then for small enough \vec{k} , $f(\vec{a} + \vec{k}) - f(\vec{a}) > 0$
 $\Rightarrow f(\vec{a} + \vec{k}) > f(\vec{a}) \Rightarrow f(\vec{a})$ is a local min

- Sps $\lambda_j < 0 \forall j \in \{1, \dots, n\}$, then $f(\vec{a} + \vec{k}) - f(\vec{a}) < 0 \Rightarrow f(\vec{a})$ is a local max

- Some $\lambda_j > 0$, others < 0

Ex: $\lambda_1 < 0, \lambda_2 > 0$ Note, $\vec{U}_1 = (1, 0, \dots, 0), \vec{U}_2 = (0, 1, 0, \dots, 0) \dots$
 $\vec{U}_i = (0, \dots, 1, \dots, 0)$ i-th index is 1
choose $\vec{k} = (t, 0, \dots, 0)$, then $f(\vec{a} + \vec{k}) - f(\vec{a}) = \frac{1}{2} \lambda_2 t^2 + R_{\vec{a}, 2}(t, 0, \dots, 0)$
 $f(\vec{a} + \vec{k}) - f(\vec{a}) > 0$ for t small and this k .
choose $\vec{k} = (0, t, 0, \dots, 0) = t \vec{U}_2$, $f(\vec{a} + \vec{k}) - f(\vec{a}) = \frac{1}{2} \lambda_2 t^2 + R_{\vec{a}, 2}(\vec{k})$ for small t
 $f(\vec{a} + \vec{k}) - f(\vec{a}) < 0 \Rightarrow$ neither a max nor min.

- Suppose same $\lambda_j = 0$

choose $\vec{k} = (0, \dots, t, \dots, 0)$
 \downarrow j-th

$$f(\vec{a} + \vec{k}) - f(\vec{a}) = 0 + R_{\vec{a}, 2}(\vec{k})$$

would then looked at higher order

$$H_{ijk} = \partial_i \partial_j \partial_k f(\vec{a})$$

For a special case of $H = \begin{pmatrix} \lambda_1 & & 0 \\ & \ddots & \\ 0 & & \lambda_n \end{pmatrix}$

General case: $\vec{k}^T H \vec{k}$
can be transformed to a special case by a change of basis.

Spectral Thm For a symmetric matrix A can diagonalize, $A = P^{-1}DP$
 $= P^TDP$

where $D = \begin{pmatrix} \lambda_1 & & 0 \\ & \ddots & \\ 0 & & \lambda_n \end{pmatrix}$ λ_j are eigenvalues and P formed

by orthonormal eigenvectors

$$\vec{k}^T A \vec{k} = \vec{k}^T P^T D P \vec{k} = (P \vec{k})^T D (P \vec{k})$$



Thm: If f is C^2 on an open subset $S \subset \mathbb{R}^n$, and $\nabla f(\vec{\alpha}) = \vec{0}$, $\vec{\alpha} \in S$. Then for $f(\vec{\alpha})$ to be a local min it is necessary that all eigenvalues of $H(\vec{\alpha})$ are nonnegative & sufficient that all eigenvalues are strictly positive.

Also, max, necessary nonpositive, sufficient strictly negative.

PROOF: Let U_1, \dots, U_n be our orthonormal basis of eigenvectors with corresponding $\lambda_1, \dots, \lambda_n$

$$\vec{k} = c_1 \vec{U}_1 + \dots + c_n \vec{U}_n$$

$$\vec{k}^T H \vec{k} = \sum_{j=1}^n \lambda_j c_j^2 \geqslant \underbrace{1}_{\text{Assume } \lambda_j > 0 \ \forall j \in \{1, \dots, n\}} \sum_{j=1}^n c_j^2 = 1 \vec{k}^T \vec{k} = 1 \vec{k} \cdot \vec{k} = 1 |\vec{k}|^2$$

$$H \vec{U}_i = \lambda_i \vec{U}_i \quad \boxed{\text{Assume } \lambda_j > 0 \ \forall j \in \{1, \dots, n\}. \exists 1 \text{ the least positive eigenvalue}}$$

$$\frac{|R_{\vec{\alpha}, 2}(\vec{k})|}{|\vec{k}|^2} < \varepsilon \text{ for } |\vec{k}| \text{ small enough}$$

$$\text{choose } \varepsilon = \frac{1}{4} \Rightarrow |R_{\vec{\alpha}, 2}(\vec{k})| \leq \frac{1|\vec{k}|^2}{4}$$

$$f(\vec{\alpha} + \vec{k}) - f(\vec{\alpha}) \geq 1 |\vec{k}|^2 - \frac{1}{4} |\vec{k}|^2 > 0 \text{ since } 1 \text{ is positive}$$

$\Rightarrow f(\vec{\alpha} + \vec{k}) > f(\vec{\alpha}) \Rightarrow \vec{\alpha}$ is a local min (here, done for sufficient case)

$$\text{Now suppose } \lambda_j < 0, f(\vec{\alpha} + t \vec{U}_j) - f(\vec{\alpha}) = \lambda_j t^2 + R_{\vec{\alpha}, 2}(t \vec{U}_j)$$

for small t get a contradiction, as $\lambda_j t^2 < 0$.

1: likewise. for $\lambda_j = 0$, so choose $\vec{k} = t \vec{U}_j$. get $f(\vec{\alpha} + \vec{k}) - f(\vec{\alpha}) = 0 + R_{\vec{\alpha}, 2}(\vec{k})$

D. Find CP's

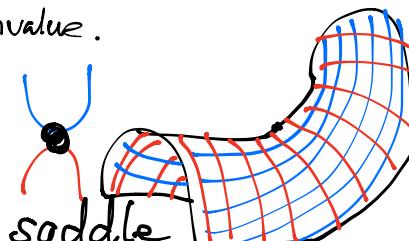
2). Compute H

3). find eigenvalues of H

4). Use Thm to determine a max/min

Defn: A saddle point has both a positive & negative eigenvalue

Def'n: A degenerate point has a zero eigenvalue.



Thm: $f : \mathbb{R}^2 \rightarrow \mathbb{R}$, C^2 on open S

$$D = \det |H(\vec{\alpha})| = \begin{vmatrix} \alpha & \beta \\ \beta & \gamma \end{vmatrix} \quad \begin{matrix} \alpha = \partial_x^2 f(\vec{\alpha}) \\ \beta = \partial_x \partial_y f(\vec{\alpha}) \\ \gamma = \partial_y^2 f(\vec{\alpha}) \end{matrix}$$

then

① $D < 0$ saddle.

② $D > 0$ and $\alpha > 0$ local min.

③ $D > 0$ & $\alpha < 0$ local max

④ $D = 0$ degenerate

Proof: $\det(H(\vec{x})) = \lambda_1 \lambda_2$

① $D < 0 \Rightarrow \lambda_1, \lambda_2$ have different sign
 \Rightarrow saddle by previous thm.

② $D > 0 \Rightarrow \lambda_1, \lambda_2$ same.

③ $\alpha = \vec{U}^T H \vec{U}$ for $\vec{U} = (1, 0)$

by possible change of basis

$d > 0 \Rightarrow \lambda_1 > 0 \Rightarrow \lambda_2 > 0 \Rightarrow$ local min
 $d < 0 \Rightarrow \lambda_1 < 0 \Rightarrow \lambda_2 < 0 \Rightarrow$ local max

④ $D = 0 \Rightarrow \alpha = 0$ eigenvalue \Rightarrow degenerate

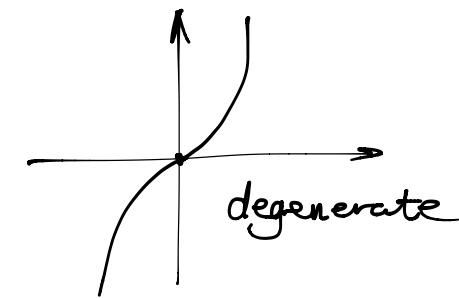
Ex: $\vec{z} = f(x, y) = x^2 + y^2$

1) find CPs

$$\nabla f(\vec{x}) = \vec{0} \Rightarrow \partial_x f(\vec{x}) = \partial x = 0$$

$$\partial_y f(\vec{x}) = \partial y = 0 \\ \text{when } (x, y) = (0, 0)$$

$$H(\vec{x}) = \begin{pmatrix} 2 & 0 \\ 0 & 2 \end{pmatrix} \quad \lambda_1 = 0 = \lambda_2 \text{ or } D = 4, d = 2 \Rightarrow \text{local min}$$



Ex: $\vec{z} = f(x, y) = (x+1)xy^2 - x^2$

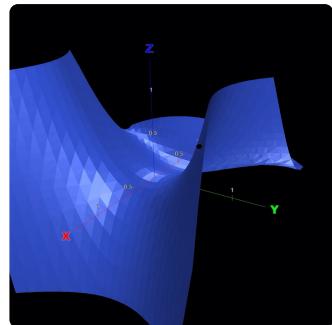
$$\begin{aligned} \partial_x f &= (y^2 - x^2) - 2x(x+1) = y^3 - 3x^2 - 2x = 0 \\ \partial_y f &= 2y(x+1) = 0 \Rightarrow y = 0 \text{ or } x = -1 \end{aligned}$$

$$\text{if } y = 0 \Rightarrow -3x^2 - 2x = 0$$

$$\Rightarrow x = 0 \text{ or } x = -2/3$$

$$\text{if } x = -1, y^2 = 1 \Rightarrow y = \pm 1$$

CPs are $(0, 0, 0)$, $(-\frac{2}{3}, 0, -\frac{4}{27})$, $(-1, 1, 0)$



and $(-1, -1, 0)$

$$\partial_x^2 f = -6x - 2$$

$$\partial_y^2 f = 2(x+1)$$

$$\partial_x \partial_y f = 2y = \partial_y \partial_x f$$

$f(x, y)$ is C^∞

$$D = -4(3x+1)(x+1) - 4y^2$$

$D|_{(0,0)} = -4 \Rightarrow$ saddle

$D|_{(-1,1)} = -4 = D|_{(-1,0)}$ two saddles

$D|_{(-\frac{2}{3}, 0)} = \frac{4}{3} > 0 \quad d > 0 \Rightarrow$ local min

$$H(\vec{x}) = \begin{pmatrix} -2(3x+1) & 2y \\ 2y & 2(x+1) \end{pmatrix}$$

