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We have Q pos. def. on V_0 since any $\bar{x} \in V_0$ has

$$Q(\bar{x}) = \alpha_1(\bar{x})^2 + \dots + \alpha_k(\bar{x})^2 = \underbrace{\alpha_{k+1}(\bar{x})^2}_0 + \dots + \underbrace{\alpha_{k+l}(\bar{x})^2}_0 \geq 0$$

with $Q(\bar{x}) = 0 \iff \alpha_1(\bar{x}) = \dots = \alpha_k(\bar{x}) = 0$

$$\iff \alpha_1(\bar{x}) = \dots = \alpha_k(\bar{x}) = \alpha_{k+1}(\bar{x}) = \dots = \alpha_n(\bar{x}) = 0$$

since $\bar{x} \in V_0 = \ker \varphi$

$$\iff \bar{x} = \bar{0}$$

since $\alpha_1, \alpha_2, \dots, \alpha_n$ are a basis for all linear functions $\mathbb{R}^n \rightarrow \mathbb{R}$ \blacksquare

2/6/2017 >

This now equips us for...

§3.6 Classifying critical points

THM 3.6.3: Given $\overset{\text{open}}{U} \xrightarrow{f} \mathbb{R}$ with f differentiable,
 \uparrow
 \mathbb{R}^n

if $\bar{x}_0 \in U$ is an extremum (max or min) for f on U

then $[Df(\bar{x}_0)] = \bar{0}$, i.e. $\frac{\partial f}{\partial x_1}(\bar{x}_0) = \dots = \frac{\partial f}{\partial x_n}(\bar{x}_0) = 0$

$$\left[\frac{\partial f}{\partial x_1}(\bar{x}_0) \dots \frac{\partial f}{\partial x_n}(\bar{x}_0) \right]$$

$$\uparrow$$

$$\nabla f(\bar{x}_0)^T$$

(in which case we call \bar{x}_0 a critical point for f
 and $f(\bar{x}_0)$ a critical value for f ; DEF'N 3.6.4)

proof: We've seen this argument before: assume the min case, and

$$\frac{\partial f}{\partial x_i}(\bar{x}_0) = \lim_{t \rightarrow 0} \frac{f(\bar{x}_0 + te_i) - f(\bar{x}_0)}{t} \text{ exists since } f \text{ is diff'ble at } \bar{x}_0$$

$$= \begin{cases} \lim_{t \rightarrow 0^+} \frac{f(\bar{x}_0 + te_i) - f(\bar{x}_0)}{t} \leftarrow \text{nonnegative} \right\} \Rightarrow \geq 0 \\ \lim_{t \rightarrow 0^-} \frac{f(\bar{x}_0 + te_i) - f(\bar{x}_0)}{t} \leftarrow \text{nonnegative} \right\} \Rightarrow \leq 0. \end{cases}$$

(t) \leftarrow positive
 (t) \leftarrow negative

so $\frac{\partial f}{\partial x_i}(\bar{x}_0) = 0$ \blacksquare

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THM 3.6.9. If $U \xrightarrow{\text{open } f} \mathbb{R}$ has a critical point $\bar{a} \in U$

and $f(\bar{a}+h) = \underbrace{f(\bar{a}) + \frac{1}{2}h^T H h}_{P^2_{f,\bar{a}}(\bar{a}+h)} + o(|h|^2)$ with $H = \left[\frac{\partial^2 f}{\partial x_i \partial x_j}(\bar{a}) \right]$
the Hessian at \bar{a}
of signature (k, l) ,

then

(i) $k=n$ (i.e. $(k, l) = (n, 0)$) $\Rightarrow f(\bar{a})$ is a local min 

(ii) $l=n$ (i.e. $(k, l) = (0, n)$) $\Rightarrow f(\bar{a})$ is a local max 

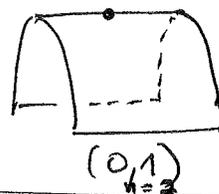
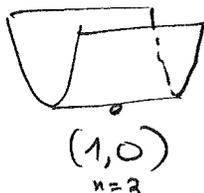
(iii) $k > 0 \Rightarrow f(\bar{a})$ is not a local max since $\exists b$ arbitrarily close to \bar{a} with $f(b) > f(\bar{a})$

(iv) $l > 0 \Rightarrow$ — " — local min — " — $f(b) < f(\bar{a})$

(and if both $k, l > 0$, $f(\bar{a})$ is called a saddle point 

REMARK: $\left\{ \begin{array}{l} (k, 0) \text{ with } k < n \\ (0, l) \text{ with } l < n \end{array} \right\}$ require further Taylor terms to analyze

\hookrightarrow called degenerate critical points



proof of THM 3.6.9: (ii) follows from (i), and (iv) from (iii), replacing f by $-f$.

Note $\frac{f(\bar{a}+h) - f(\bar{a})}{|h|^2} = \frac{\frac{1}{2}h^T H h}{|h|^2} + \underbrace{\frac{o(|h|^2)}{|h|^2}}_{\rightarrow 0 \text{ as } h \rightarrow 0}$

So to prove (i), it's enough to check $\lim_{h \rightarrow 0} \frac{h^T H h}{|h|^2} > C$ for some positive constant $C > 0$.

In fact, if H has orthonormal eigenbasis $\bar{v}_1, \dots, \bar{v}_n$ with $0 < \lambda_1 \leq \dots \leq \lambda_n$ as eigenvalues \hookrightarrow since $k=n$

then we'll show $\frac{h^T H h}{|h|^2} > \lambda_1$:

(22) Write $\bar{h} = c_1 \bar{v}_1 + \dots + c_n \bar{v}_n$ for some $c_i \in \mathbb{R}$.

Then $|\bar{h}|^2 = \bar{h}^T \bar{h} = \left(\sum_{i=1}^n c_i \bar{v}_i \right)^T \left(\sum_{i=1}^n c_i \bar{v}_i \right) = c_1^2 + c_2^2 + \dots + c_n^2$
 because $\bar{v}_i^T \bar{v}_j = \begin{cases} 1 & \text{if } i=j \\ 0 & \text{if } i \neq j \end{cases}$

while $\bar{h}^T H \bar{h} = \left(\sum_{i=1}^n c_i \bar{v}_i \right)^T \left(\sum_{i=1}^n c_i H \bar{v}_i \right)$

$= \left(\sum_{i=1}^n c_i \bar{v}_i \right)^T \left(\sum_{i=1}^n c_i \lambda_i \bar{v}_i \right)$

$= \lambda_1 c_1^2 + \lambda_2 c_2^2 + \dots + \lambda_n c_n^2$

$\geq \lambda_1 c_1^2 + \lambda_1 c_2^2 + \dots + \lambda_1 c_n^2 = \lambda_1 (c_1^2 + \dots + c_n^2)$

$\lambda_1 \leq \lambda_2 \leq \dots \leq \lambda_n$

$\Rightarrow \frac{\bar{h}^T H \bar{h}}{\bar{h}^T \bar{h}} = \frac{\lambda_1 \left(\sum_{i=1}^n c_i^2 \right)}{\sum_{i=1}^n c_i^2} = \lambda_1 (> 0).$

To prove (ii), it's enough to check that if $H\bar{v} = \lambda\bar{v}$ with $\lambda > 0$
 (and such a \bar{v} exists since $k > 0$)

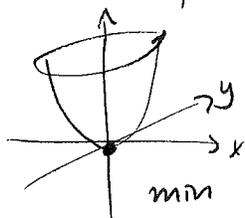
then taking $\bar{h} = t\bar{v}$, $\frac{\bar{h}^T H \bar{h}}{|\bar{h}|^2} > 0$

~~...~~ $\frac{(t\bar{v})^T H (t\bar{v})}{|t\bar{v}|^2}$

~~...~~ $= \frac{(t\bar{v})^T \lambda t\bar{v}}{|t\bar{v}|^2} = \lambda \frac{(t\bar{v})^T \bar{v}}{|t\bar{v}|^2} = \lambda (> 0) \quad \blacksquare$

^(Book's) EXAMPLES of degenerate critical points all with $P_{f,0} = x^2$ and $n=2$, signature of H being $(1,0)$:

$f(x,y) = x^2 + y^4$

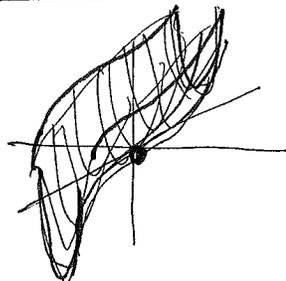


$f(x,y) = x^2 - y^4$



saddle

$f(x,y) = x^2 - y^3$



see book's much better plot!

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§3.7 Constrained extrema & Lagrange multipliers

How to find the points \bar{c} on a k -dimensional manifold $M \subset \mathbb{R}^n$ that give (local) maxes/mins for some $\mathbb{R}^n \xrightarrow{f} \mathbb{R}$?

If we have a local parametrization $U \xrightarrow{\bar{\gamma}} \mathbb{R}^n$ of M ,

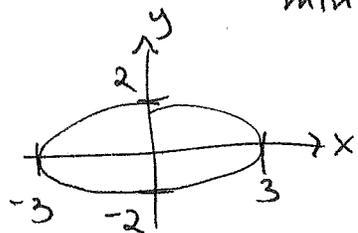
$$\begin{array}{c} \cap \\ \mathbb{R}^k \\ \bar{t} = \begin{pmatrix} t_1 \\ \vdots \\ t_k \end{pmatrix} \end{array} \mapsto \bar{\gamma} \begin{pmatrix} t_1 \\ \vdots \\ t_k \end{pmatrix} = \begin{pmatrix} x_1 \\ \vdots \\ x_n \end{pmatrix} = \begin{pmatrix} \gamma_1(\bar{t}) \\ \vdots \\ \gamma_n(\bar{t}) \end{pmatrix}$$

just find the critical points in \mathbb{R}^k for the composite function $f \circ \bar{\gamma}$:

$$\begin{array}{c} U \\ \cap \\ \mathbb{R}^k \\ \bar{t} \end{array} \xrightarrow{\bar{\gamma}} \mathbb{R}^n \xrightarrow{f} \mathbb{R}$$

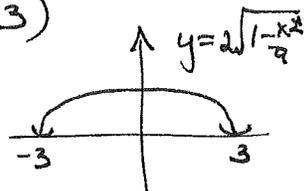
$$\bar{t} \longmapsto \bar{\gamma}(\bar{t}) \longmapsto f \begin{pmatrix} \gamma_1(\bar{t}) \\ \vdots \\ \gamma_n(\bar{t}) \end{pmatrix}$$

EXAMPLE: What points on the ellipse $\frac{x^2}{9} + \frac{y^2}{4} = 1$ minimize and maximize $f \begin{pmatrix} x \\ y \end{pmatrix} = x + y$?



We can parametrize on this $U \subset (-3, 3)$

$$\text{via } y = +2\sqrt{1 - \frac{x^2}{9}} = g(x)$$



$$\begin{aligned} \text{where } f \begin{pmatrix} x \\ y \end{pmatrix} &= f \begin{pmatrix} x \\ g(x) \end{pmatrix} = x + g(x) \\ &= x + 2\sqrt{1 - \frac{x^2}{9}} \end{aligned}$$

so the critical points are where

$$0 = \frac{d}{dx} \left(x + 2\sqrt{1 - \frac{x^2}{9}} \right) = 1 + 2 \left(\frac{1}{2} \right) \left(-\frac{2x}{9} \right) \left(1 - \frac{x^2}{9} \right)^{-\frac{1}{2}}$$

$$0 = 1 - \frac{2x}{9 \left(1 - \frac{x^2}{9} \right)^{\frac{1}{2}}}$$

$$\text{square } \left(\frac{2x}{9 \left(1 - \frac{x^2}{9} \right)^{\frac{1}{2}}} \right)^2 = 1$$

$$\frac{4x^2}{81 \left(1 - \frac{x^2}{9} \right)} = 1$$

$$\begin{aligned} 4x^2 &= 81 - 9x^2 \\ 13x^2 &= 81, \quad x = \pm \frac{9}{\sqrt{13}} \end{aligned}$$

