

Higher Partial derivatives and the Hessian.

Let $f(x,y) = \begin{cases} xy & (x,y) \neq (0,0) \\ 0 & (x,y) = (0,0) \end{cases}$

then $\frac{\partial f}{\partial x}(0,0) \neq \frac{\partial f}{\partial y}(0,0)$

A sufficient condition for the equality of mixed partial derivatives is

Thm let $f \in C^2(\Omega)$ $f: \Omega \rightarrow \mathbb{R}$, $\Omega \subset \mathbb{R}^n$
(i.e $\frac{\partial f}{\partial x_i}$, $\frac{\partial f}{\partial x_i x_j}$ are continuous). Then

$$\frac{\partial^2 f}{\partial x_i \partial x_j} = \frac{\partial^2 f}{\partial x_j \partial x_i} \quad 1 \leq i, j \leq n.$$

Defn let $f: \mathcal{O} \subset \mathbb{R}^n \rightarrow \mathbb{R}$

The $n \times n$ Matrix $H(f; x_0) = (D_{ij} f(x_0))_{i,j=1}^n$,

the matrix whose (i,j) entry is $\frac{\partial^2 f}{\partial x_i \partial x_j}(x_0)$,

is called the Hessian Matrix

First order Taylor formulae

$f: \mathcal{O} \rightarrow \mathbb{R}$, $\mathcal{O} \subset \mathbb{R}^n$. f is differentiable at x_0 . Then

$$f(x) = f(x_0) + \nabla f(x_0) \cdot (x - x_0) + \underbrace{R(|x - x_0|)}_{O(|x - x_0|)}$$

where

$$\lim_{x \rightarrow x_0} \left(\frac{R(x; x_0)}{|x - x_0|} \right) = 0$$

The polynomial $T_1(x) = f(x_0) + \nabla f(x_0) \cdot (x - x_0)$ is the first order approximation to f near x_0 . (2)

Thm. Let $f : \mathbb{R}^n \rightarrow \mathbb{R}$ with continuous second order partial derivatives. Then $\forall y \in \mathbb{R}^n$

$$f(x+y) = f(x) + \nabla f(x) \cdot y + \frac{1}{2!} y^T H_f(x) y + R_2(x, y)$$

where $\frac{R_2(x, y)}{\|x-y\|^2} \rightarrow 0$ as $x \rightarrow y$.

or equivalently

$$\begin{aligned} f(x) &= f(x_0) + \nabla f(x_0) \cdot (x-x_0) + \frac{1}{2} (x-x_0)^T H_f(x_0) (x-x_0) + \\ &\quad + R_2(x, x_0) \end{aligned}$$

where $\frac{R_2(x, x_0)}{\|x-x_0\|^2} \rightarrow 0$ as $x \rightarrow x_0$.

(3)

$$f(x_1', x_2') = f(x_0', x_0'') + \sum_{i=1}^n \frac{\partial f}{\partial x_i} (x_i' - x_i'') +$$

$$\frac{1}{2} \sum_{i=1}^n \sum_{j=1}^n \frac{\partial^2 f}{\partial x_i \partial x_j} (x_0) (x_i'' - x_i') (x_j'' - x_j').$$

Rk. Sometimes Hessian Matrix is also denoted by

$$\nabla^2 f$$

There is a general Taylor formula which looks as follows.

Thm Let $f \in C^k(\mathcal{U})$, $x, x_0 \in \mathcal{U}$. Then

$$f(x) = f(x_0) + Df(x_0) \cdot (x - x_0) + \frac{1}{2} \sum_{i,j=1}^n \frac{\partial^2 f}{\partial x^i \partial x^j} (x^i - x_0^i)(x^j - x_0^j)$$

$$\dots + \frac{1}{k!} \sum_{i_1, i_2, \dots, i_k=1}^n \frac{\partial^k f}{\partial x^{i_1} \partial x^{i_2} \dots \partial x^{i_k}} (x^{i_1} - x_0^{i_1})(x^{i_2} - x_0^{i_2}) \dots (x^{i_k} - x_0^{i_k}) + R_k(x, x_0)$$

such that

$$\lim_{||x-x_0||^k} \frac{R_k(x-x_0)}{||x-x_0||^k} = 0 \quad \text{as } x \rightarrow x_0.$$

$$\text{Given: } f(x,y) = e^{x+y} \cos x$$

at pt $(0,0)$

at pt $(0,0)$

Find the Taylor expansion of degree 3 around $(0,0)$

$$f(0,0) = 1.$$

$$\frac{\partial f}{\partial x} = e^x e^y \cos x - e^{x+y} \sin x \Big|_{(0,0)} = 1.$$

$$\frac{\partial f}{\partial y} = e^x \cos x e^y \Big|_{(0,0)} = 1.$$

$$\frac{\partial^2 f}{\partial x^2} = e^y [e^x (\cos x - \sin x) + e^x (-\sin x - \cos x)] \Big|_{(0,0)} = 0.$$

$$\frac{\partial^2 f}{\partial y^2} = e^y e^x (\cos x - \sin x) \Big|_{(0,0)} = 0.$$

$$f_x(0,0) = 1.$$

$$f_y(0,0) = 1.$$

$$\left. \frac{\partial^3 f}{\partial x^3} = -2e^y (e^x \sin x + e^x \cos x) \right|_{(0,0)} = -2$$

$$\left. \frac{\partial^3 f}{\partial y^3} = e^y e^x \cos x \right|_{(0,0)} = 1.$$

$$\left. \frac{\partial^3 f}{\partial x^2 \partial y} = e^y e^x (2 \sin x) \right|_{(0,0)} = 0 = \left. \frac{\partial^3 f}{\partial x \partial y^2} \right|_{(0,0)}$$

$$\left. \frac{\partial^3 f}{\partial x^3} \right|_{(0,0)} = 1.$$

$$\left. \frac{\partial^3 f}{\partial x^2 \partial y} \right|_{(0,0)}$$

Hence

$$e^{xy} \cos x = 1 + \left. \nabla f(0,0) \cdot \vec{R}(x,y) \right|_{(0,0)} + \frac{\partial^2 f}{\partial x^2} \left(\frac{x^2}{2} y^2 \right) + \frac{\partial^2 f}{\partial y^2} \left(\frac{y^2}{2} x^2 \right)$$

$$= 1 + x + y + \frac{1}{2} \left((-2)x^3 + (3)xy^2 + y^3 \right)$$

$$+ \frac{1}{2} \left(\frac{\partial^2 f}{\partial x^3} \left(x^3 \right) + \frac{\partial^2 f}{\partial y^3} \left(y^3 \right) \right).$$

$$T_3(x,y) = 1 + x + y + \frac{1}{2} (2xy + y^2) + \frac{1}{6} (-2x^3 + 3xy^2 + y^3).$$

$$e^x = \sum_{k=0}^{\infty} \frac{x^k}{k!}$$

conv

for all x .

$$\frac{1}{1-x} = \sum x^k ,$$

$$|x| < 1$$

$$\frac{1}{1-y^2} = \sum y^{2k} , \quad |y^2| < 1$$

(3)

lokale Extremae von Funktionen $f: \mathbb{R}^n \rightarrow \mathbb{R}$.

$$f: \mathbb{R}^n \rightarrow \mathbb{R}$$

für $f: \mathbb{R} \rightarrow \mathbb{R}$ we have seen that local

extrema of f occur at critical points

(i.e. points for which $f'(x) = 0$)

The nature of extrema is determined by $f''(x)$.
If x_0 is a critical pt., then

(i) x_0 is a local min if $f''(x_0) > 0$

(ii) x_0 " max if $f''(x_0) < 0$

(iii) x_0 is a saddle point if $f''(x_0) = 0$.

für $f: \mathbb{R}^n \rightarrow \mathbb{R}$, the role of $f'(x)$ is taken over by ∇f
 $f''(x)$ is " " by $H(f)$.

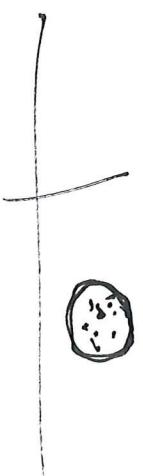
Defn Let $f: \cup \subset \mathbb{R}^n \rightarrow \mathbb{R}$ a scalar field.

A point $a = (a^1, a^2, \dots, a^n)$ is a local maximum (resp local min.) if we can find a neighborhood

$B_a(r) := \{x \in \mathbb{R}^n \mid \|x-a\| < r\}$ completely contained

such that $\forall x \in B_a(r) \quad f(x) \leq f(a)$

(resp. $f(x) \geq f(a)$).



A local extreme is a point $a \in \cup$ which is a local maximum or local minimum.

Thm. If f is a differentiable scalar field

and $a = (a^1 \dots a^n)$ is a local extremum

then $\nabla f(a) = 0$.

Defn A stationary point or a critical point

of $f: \Omega \subset \mathbb{R}^n \rightarrow \mathbb{R}$ is a point $a \in \Omega$

where $\nabla f(a) = 0$.

It is called a saddle point if it is not
a local extreme

Rk- If a is a saddle point then every nbhd $B_a(r)$ contains points x with $f(x) \geq f(a)$ and points x' such that $f(x') \leq f(a)$.

$$\underline{\underline{Ex}}: f(x,y) = x^2 - y^2$$

$$\nabla f(x,y) = (2x, -2y)$$

$$\nabla f(x,y) = (0,0) = (2x, -2y) \implies x=0, y=0.$$

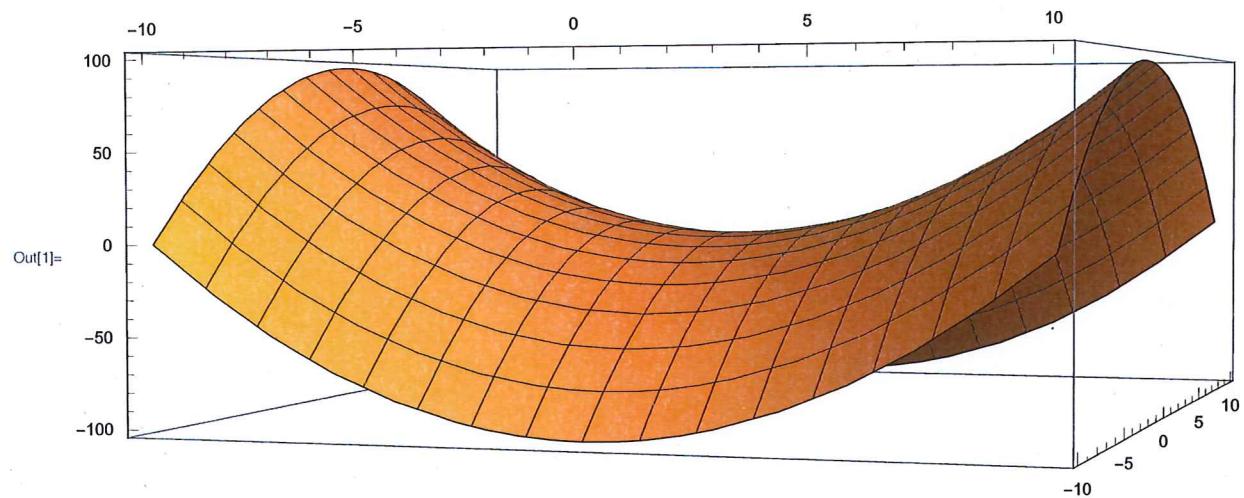
$(0,0)$ is the only critical point

It is a saddle point. Because: if we move away from it along the y -axis the function

$$is f(0,y) = -y^2 < 0 = f(0,0)$$

If we move along the x -axis, the function is $f(x,0) = x^2 \geq 0 = f(0,0)$

In[1]:= Plot3D[x^2 - y^2, {x, -10, 10}, {y, -10, 10}]



To classify critical points we will use the

Hessian Matrix.

Assume $f \in C^2$ so that 2nd order mixed partials are equal and hence the Hessian Matrix is symmetric.

Defn. A symmetric Matrix $A = (a_{ij}) \in \mathbb{R}^{n \times n}$ is called

1) positive definite

$$\forall x \in (\mathbb{R}^n \setminus \{0\}) \quad x^T A x = \sum a_{ij} x_i x_j > 0$$

If all its eigenvalues $\lambda_1, \dots, \lambda_n > 0$.

2) negative definite

$$x^T A x < 0 \quad \forall x \in \mathbb{R}^n \setminus \{0\}$$

$$\lambda_1, \dots, \lambda_n < 0.$$

3) otherwise it is called indefinite (it has both positive and negative eigenvalues or zero ev.).

In the case $n=2$ of 2×2 matrices, the (in)definiteness is easily determined using the following theorem.

Theorem A ~~symmetric~~ 2×2 Matrix $A = \begin{pmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{pmatrix}$

with $\det A \neq 0$ is

- 1) positive definite $\Leftrightarrow a_{11} > 0$ and $\det A > 0$
- 2) negative definite $\Leftrightarrow a_{11} < 0$ and $\det A > 0$
- 3) indefinite $\Leftrightarrow \det A < 0$.

$$\begin{pmatrix} m_{11} & m_{12} \\ m_{21} & m_{22} \end{pmatrix} \cdot \begin{pmatrix} m_{11} \\ m_{21} \end{pmatrix} = \begin{pmatrix} m_{11} \\ m_{21} \end{pmatrix}$$

For the local extrema of $f: \mathbb{R}^n \rightarrow \mathbb{R}$.

Thm Let $f: \mathcal{D} \rightarrow \mathbb{R}$ $\mathcal{D} \subset \mathbb{R}^n$ be

a scalar field. Assume $f \in C^2$. Let
 \mathbf{a} be a critical point of f ($\nabla f(\mathbf{a}) = \mathbf{0}$)

with $\det(\text{Hess } f(\mathbf{a})) \neq 0$

1) If $\text{Hess}(f(\mathbf{a})) > 0$ (positive definite) then \mathbf{a} is a local minimum.

2) If $\text{Hess}(f(\mathbf{a})) < 0$ (negative def.) then \mathbf{a} is a local maximum.

3) otherwise it is a saddle point.

Note If $\det(\text{Hess}) = 0$, i.e. Hess is a singular Matrix
the test does not say anything.

In the special case $f: \mathbb{R}^2 \rightarrow \mathbb{R}$ we have

Thm If $f \in C^2$, a is a critical point ($f'(x,y)$)

then 1) If $\frac{\partial^2 f}{\partial x^2}(a) < 0$ and $\det(\text{Hess } f(a)) > 0$

then a is a local max of f .

2) If $\frac{\partial^2 f}{\partial x^2}(a) > 0$, $\det(\text{Hess } f) > 0$.

then a is a local min of f

3) If $\det(\text{Hess } f(a)) < 0$ then a is a saddle point.

E_x.

1) $f(x, y, z) = (x-1)^2 + (y+2)^2 + (z+1)^2$

$$\nabla f = (2(x-1), 2(y+2), 2(z+1)) = (0, 0, 0)$$

$\Rightarrow x=1$ $\Rightarrow (1, -2, -1)$ is the only critical point.

$$\begin{cases} y=-2 \\ z=-1 \end{cases}$$

$$\text{Hess}(f) =$$

$$\begin{pmatrix} 2 & 0 & 0 \\ 0 & 2 & 0 \\ 0 & 0 & 2 \end{pmatrix}$$

$\lambda_0 > 0$ positive definite

$$(1, -2, -1)$$

is a local min.

$$\textcircled{2} \quad f(x,y) = \cos(x+2y) + \cos(2x+3y)$$

$$\nabla f = (-\sin(x+2y) - 2\sin(2x+3y), -2\sin(x+2y) - 3\sin(2x+3y))$$

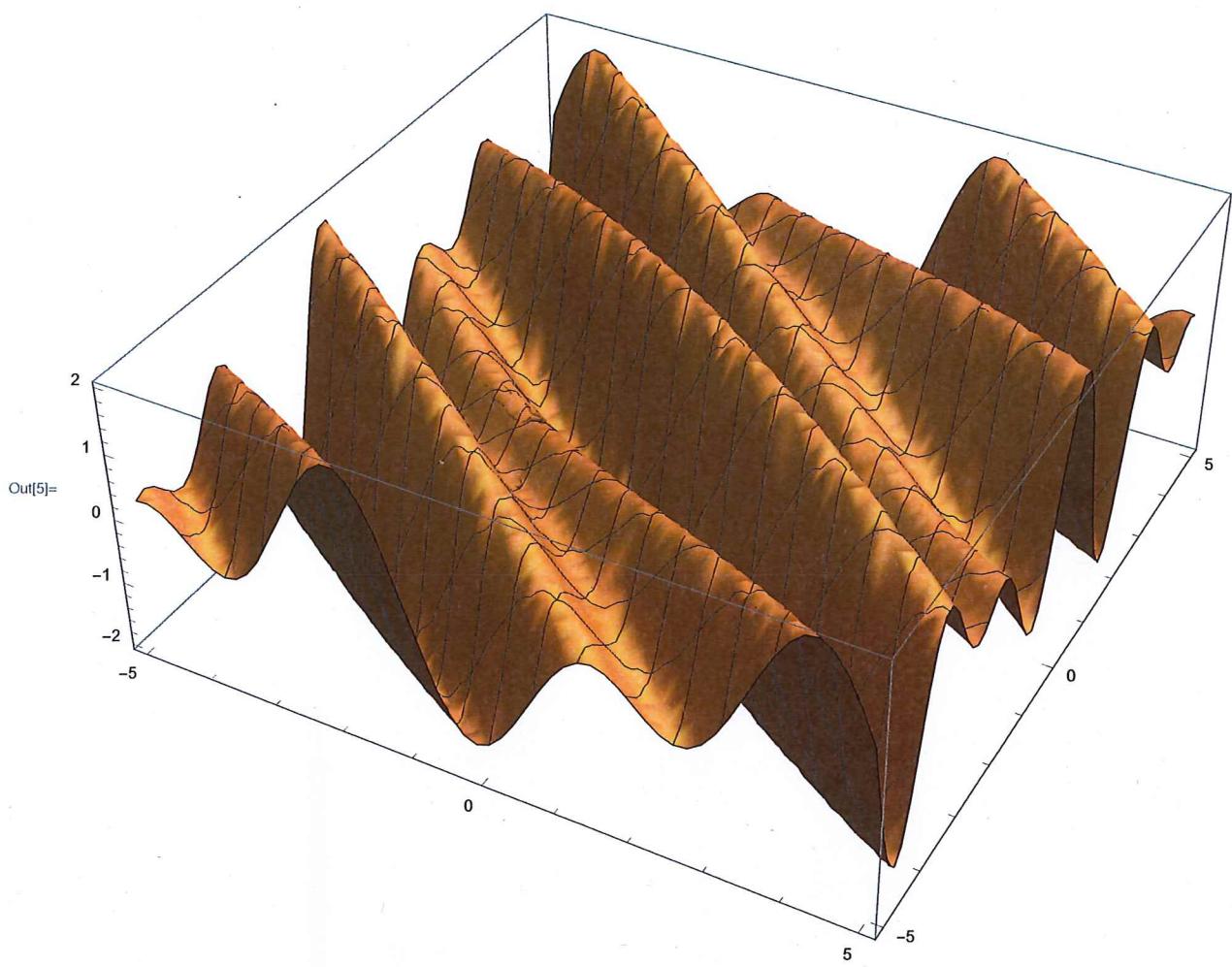
$$= (0, 0)$$

$$\Rightarrow -\sin(x+2y) - 2\sin(2x+3y) = 0 \quad \left. \begin{array}{l} \sin(2x+3y) = 0 \\ \sin(x+2y) = 0 \end{array} \right\} \Rightarrow$$

$$\Rightarrow x+2y = k\pi \quad \Rightarrow x \in k\pi, y \in \ell\pi \\ 2x+3y = \ell\pi \quad \text{with } k, \ell \in \mathbb{Z}.$$

$$\text{critical pts} = \{(x_k, y_\ell) \mid \ell, k \in \mathbb{Z}\}.$$

In[5]:= Plot3D[Cos[x + 2 y] + Cos[2 x + 3 y], {x, -5, 5}, {y, -5, 5}]



(19 L
e)

$$\frac{\partial^2 f}{\partial x^2} = -2 \cos(x+2y) - 6 \cos(2x+3y)$$

$$\frac{\partial^2 f}{\partial x^2} = -\cos(x+2y) - 4 \cos(2x+3y)$$

$$\frac{\partial^2 f}{\partial y^2} = -4 \cos(x+2y) - 9 \cos(2x+3y)$$

at pt $(0,0)$ $\nabla^2 f(0,0) = \begin{pmatrix} -5 & -8 \\ -8 & -13 \end{pmatrix}$ $\det = 1 > 0.$

$$\frac{\partial^2 f}{\partial x^2}(0,0) = -5 < 0$$

Hence $(0,0)$ is a local maximum.

Similarly for all points $(2\pi k, 2\pi \ell)$ by periodicity.

$$\text{Analog} = \text{at } (\pi, \pi) \quad \frac{\partial^2 f}{\partial x^2} = 5 > 0$$

$$\frac{\partial^2 f}{\partial y^2} = 13$$

$$\frac{\partial^2 f}{\partial x \partial y} = 8, \quad \text{hess}(f) = \begin{pmatrix} 5 & 8 \\ 8 & 13 \end{pmatrix} > 0$$

$$\det > 0$$

$$\frac{\partial^2 f}{\partial x^2} > 0$$

(π, π) is a loc min.

Similar by permuting also $((2k+1)\pi, (2l+1)\pi)$
are all loc minima.

Check: $(0, \pi)$ and $(\pi, 0)$ Hence also
 $(2k\pi, (2l+1)\pi)$ $((2k+1)\pi, 2l\pi)$ are saddle points.

Global Extreme.

Recall

$$f: [a, b] \rightarrow \mathbb{R}$$

we saw that
the global extreme of f is either at a
critical point inside (a, b) or at the boundary
points.



Then let f be differentiable in the interior of Ω .
Then every global extremum of f is either
at a critical point in the interior of Ω or
on the boundary of Ω .

