

# SE102:Multivariable Calculus

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Lecture 04  
Maxima and Minima

## Definition

An **open ball**  $B_\varepsilon(\mathbf{a})$  of the radius  $\varepsilon > 0$  at a point  $\mathbf{a} \in \mathbf{R}^n$  is the set defined by

$$B_\varepsilon(\mathbf{a}) = \{\mathbf{x} \in \mathbf{R}^n \mid \|\mathbf{x} - \mathbf{a}\| < \varepsilon\}$$

## Definition

A function  $f(x_1, \dots, x_n) = (y_1, \dots, y_m)$  is said to be **continuously differentiable**, or of **class  $\mathcal{C}^1$** , at  $\mathbf{a} \in \mathbf{R}^n$  if there is  $\varepsilon > 0$  such that all the partial derivatives  $\frac{\partial y_j}{\partial x_i}$  are continuous on an open ball  $B_\varepsilon(\mathbf{a})$ .

## Theorem

*If a function  $z = f(x, y)$  is continuously differentiable at  $\mathbf{a} = (x_0, y_0)$ , then it is differentiable at  $\mathbf{a}$ .*

## Proof.

Let  $B_\varepsilon(\mathbf{a})$  be an open ball on which the partial derivatives  $f_x, f_y$  are continuous. Let us choose a point  $\mathbf{b} = (x, y) \in B_\varepsilon(\mathbf{a})$  and define a sequence

$$\mathbf{p}_0 = \mathbf{a}, \quad \mathbf{p}_1 = (x, y_0), \quad \mathbf{p}_2 = \mathbf{b}$$

Without loss of generality, we will assume that  $x_0 \leq x$  and  $y_0 \leq y$ . Define a function  $\phi_1(t) = f(t, y_0)$  on  $[x_0, x]$ . Note that  $\phi_1'(t) = f_x(t, y_0)$ . By the mean value theorem on one-variable function, there is  $t_1 \in [x_0, x]$  such that

$$\phi_1'(t_0)(x - x_0) = f(x, y_0) - f(x_0, y_0)$$

### Proof.

Likewise, let us define  $\phi_2(t) = f(x, t)$  on  $[y_0, y]$ , then there exists  $t_2 \in [y_0, y]$  satisfying

$$\phi_2'(t_2)(y - y_0) = f(x, y) - f(x, y_0)$$

Then,

$$\begin{aligned} f(\mathbf{b}) - f(\mathbf{a}) - f_x(\mathbf{a})(x - x_0) - f_y(\mathbf{a})(y - y_0) \\ &= (\phi_1'(t_1) - f_x(\mathbf{a}))(x - x_0) + (\phi_2'(t_2) - f_y(\mathbf{a}))(y - y_0) \\ &= (f_x(t_1, y_0) - f_x(\mathbf{a}))(x - x_0) + (f_y(x, t_2) - f_y(\mathbf{a}))(y - y_0) \end{aligned}$$

Note that as  $\mathbf{b} \rightarrow \mathbf{a}$ , we have  $t_1 \rightarrow x_0$  and  $t_2 \rightarrow y_0$ , which means the above goes to 0. Since  $(x - x_0)/\|\mathbf{b} - \mathbf{a}\|$  and  $(y - y_0)/\|\mathbf{b} - \mathbf{a}\|$  bounded, we proved the differentiability of  $f$  at  $\mathbf{a}$ .

## Definition

A function  $z = f(x, y)$  is said to be of **class**  $\mathcal{C}^2$  at  $\mathbf{a} \in \mathbf{R}^2$  if there exists  $\varepsilon > 0$  such that all second-order partial derivatives are continuous on an open ball  $B_\varepsilon(\mathbf{a})$ .

## Remark

Given a function  $\mathbf{f} : \mathbf{R}^n \rightarrow \mathbf{R}^m$ , the derivative  $\mathbf{Df}$  is a function  $\mathbf{Df} : \mathbf{R}^n \rightarrow \mathbf{R}^{nm}$  whose coordinate functions are all the first derivatives of  $\mathbf{f}$ . The function  $\mathbf{f}$  being a class  $\mathcal{C}^1$  means that  $\mathcal{D}\{\mathbf{f}\}$  is continuous. Similarly, being a class  $\mathcal{C}^2$  means that  $\mathbf{Df}$  is  $\mathcal{C}^1$ , which equivalent to saying that  $\mathbf{D}^2\mathbf{f}$  is continuous. We say the function  $\mathbf{f}$  is of class  $\mathcal{C}^n$  if  $\mathbf{D}^n\mathbf{f}$  is continuous. Moreover, the function  $\mathbf{f}$  is of class  $\mathcal{C}^\infty$  if  $\mathbf{D}^n\mathbf{f}$  is continuous for all  $n > 0$ .

## Theorem (Clairaut)

If  $z = f(x, y)$  is of class  $\mathcal{C}^2$  at  $(x_0, y_0)$ , then

$$f_{xy}(x_0, y_0) = f_{yx}(x_0, y_0).$$

## Example

Check if the following function satisfies Clairaut's theorem at  $(0, 0)$ .

$$f(x, y) = \begin{cases} xy \frac{x^2 - y^2}{x^2 + y^2} & (x, y) \neq (0, 0) \\ 0 & (x, y) = (0, 0) \end{cases}$$

## Proof of Clairaut's Theorem.

Let us define  $\mu : B_\varepsilon(\mathbf{a}) \rightarrow \mathbf{R}$  as

$$\mu(x, y) = f(x, y) - f(x_0, y) - f(x, y_0) + f(x_0, y_0)$$

Let us define  $\phi(t) = f(t, y) - f(t, y_0)$  on  $[x_0, x]$ . Since  $f$  is of class  $\mathcal{C}^1$ ,  $\phi$  is differentiable on  $[x_0, x]$ . Note that

$$\mu(x, y) = \phi(x) - \phi(x_0).$$

By the mean value theorem, there exists  $t_1 \in [x_0, x]$  satisfying

$$\mu(x, y) = \phi'(t_1)(x - x_0) = (f_x(t_1, y) - f_x(t_1, y_0))(x - x_0).$$

Since  $f_x$  is of class  $\mathcal{C}^1$ , we can apply the mean value theorem on  $\psi(h) = f_x(t_1, h)$ , which means that there exists  $t_2 \in [y_0, y]$  satisfying

$$f_x(t_1, y) - f_x(t_1, y_0) = \psi'(t_2)(y - y_0) = f_{xy}(t_1, t_2)(y - y_0)$$

### Proof.

Thus we showed that there exists  $(t_1, t_2) \in [x_0, x] \times [y_0, y]$  satisfying

$$\mu(x, y) = f_{xy}(t_1, t_2)(x - x_0)(y - y_0).$$

If we switch the role of  $x$  and  $y$  when we defined  $\phi$ , we would obtain the similar result: there exists a point  $(s_1, s_2) \in [x_0, x] \times [y_0, y]$  satisfying

$$\mu(x, y) = f_{yx}(s_1, s_2)(x - x_0)(y - y_0).$$

Let us replace  $x = x_0 + h$  and  $y = y_0 + h$ . Since we assume that  $f$  is of class  $\mathcal{C}^2$ , the following function converges as  $(x, y) \rightarrow (x_0, y_0)$ .

$$\frac{\mu(x_0 + h, y_0 + h)}{h^2} = f_{xy}(t_1, t_2) = f_{yx}(s_1, s_2)$$

Since  $(t_1, t_2), (s_1, s_2) \rightarrow (x_0, y_0)$  as  $(x, y) \rightarrow (x_0, y_0)$ , we are done.



## Definition

Suppose that  $f(x, y)$  is of class  $\mathcal{C}^2$  at  $(x_0, y_0)$ . Then the following polynomial is called the **Taylor polynomial of second degree 2** of  $f$  at  $(x_0, y_0)$ .

$$\begin{aligned} Q(x, y) = & f(x_0, y_0) + f_x(x_0, y_0)(x - x_0) \\ & + f_y(x_0, y_0)(y - y_0) + \frac{1}{2}f_{xx}(x_0, y_0)(x - x_0)^2 \\ & + f_{xy}(x_0, y_0)(x - x_0)(y - y_0) \\ & + \frac{1}{2}f_{yy}(x_0, y_0)(y - y_0)^2. \end{aligned}$$

## Remark

If all  $n$ -th order partial derivatives of a function  $f(x, y)$  are continuous, then

- ▶  $f$  is differentiable, and
- ▶ the formulae of  $n$ -order partial derivatives does not depend on the order of partial derivatives.

## Remark

Let

$$\Delta \mathbf{x} = (\Delta x, \Delta y) = (x - x_0, y - y_0)$$

and denote the *gradient operator*  $\nabla$  in vector notation:

$$\nabla = \left( \frac{\partial}{\partial x}, \frac{\partial}{\partial y} \right)$$

Let us define the *multiplication* of differential operators as follows:

$$\left(\frac{\partial}{\partial x} \frac{\partial}{\partial x}\right) f = \frac{\partial^2 f}{\partial x^2}, \quad \left(\frac{\partial}{\partial x} \frac{\partial}{\partial y}\right) f = \frac{\partial^2 f}{\partial x \partial y}$$

By the assumption in the definition of Taylor polynomial, we have  $f_{xy}(x_0, y_0) = f_{yx}(x_0, y_0)$ . Thus we can write

$$Q(x, y) = \sum_{n=0}^2 \frac{1}{n!} (\Delta \mathbf{x} \cdot \nabla)^n f(x_0, y_0)$$

We can generalize this to the  $k$ -th order Taylor polynomial.

$$P_k f(\mathbf{x}) = \sum_{n=0}^k \frac{1}{n!} (\Delta \mathbf{x} \cdot \nabla)^n f(\mathbf{x}_0)$$

This is a generalization of the Taylor polynomial for single variable function:

## Theorem (Taylor)

Let  $z = f(x, y)$  be a function of class  $C^3$  on a rectangular region

$$D = \{(x, y) \mid |x - x_0|, |y - y_0| \leq \epsilon\}.$$

Then for each  $(x, y) \in D$ , there exists a constant  $0 \leq c \leq 1$  satisfying

$$f(x, y) = Q(x, y) + R_2(x, y)$$

where

$$R_2(x, y) = \frac{1}{3!}(\Delta \mathbf{x} \cdot \nabla)^3 f(\mathbf{x}_0 + c\Delta \mathbf{x})$$

## Remark

This theorem is a generalization of Taylor theorem for one-variable function:

*Let  $f$  be of class  $\mathcal{C}^{k+1}$  on an interval  $I = (x_0 - \epsilon, x_0 + \epsilon)$ . Then for  $x, c \in I$ , there exists a constant  $\xi$  between  $x$  and  $c$  such that*

$$f(x) = P_k(x) + \frac{f^{(k+1)}(\xi)}{(k+1)!}(x-c)^{k+1}$$

Note that the choice of  $c$  depends on the choice of  $x, x_0$ . Since all the third-order partial derivatives of  $f$  are bounded, the error  $|R_n(x, y)|$  decreases to zero, as  $(x, y) \rightarrow (x_0, y_0)$ ,

## Example

Find  $Q_2(x, y)$  at  $(0, 0)$  for

▶  $f(x, y) = xy - x^2 - 5y^2 + y - 1$

▶  $f(x, y) = \cos x \cos y$

and compare the graphs of  $Q_2$  and  $f$  near  $(1, 0)$

## Definition

Let  $z = f(x, y)$  be a function defined on an open ball at  $\mathbf{a} = (x_0, y_0)$ . The point  $\mathbf{a}$  is said to be **local maximal** (**minimal**, respectively) if there exists a sufficiently small  $\varepsilon > 0$  such that for all  $\mathbf{x} \in B_\varepsilon(\mathbf{a})$ ,

$$f(\mathbf{a}) \geq f(\mathbf{x}) \quad (f(\mathbf{a}) \leq f(\mathbf{x}), \text{ respectively}).$$

A local maximal or minimal is called an **extremal**.

## Definition

A point  $(\mathbf{a})$  is called a **critical point** if it satisfies one of the following conditions:

1.  $f_x(\mathbf{a}) = f_y(\mathbf{a}) = 0$ ;
2.  $f_x$  or  $f_y$  does not exist at  $\mathbf{a}$ ;
3.  $f$  is discontinuous at  $\mathbf{a}$ .

A critical point which is *not* an extremal point is called a **saddle point**.

## Example

Find the critical points of

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

and classify them. Also, find  $Q_2(x, y)$  at each critical points and compare their graphs.



## Remark

1. Suppose that  $f(x, y)$  is differentiable at  $\mathbf{a}$ . The linear approximation  $Q_1$  of  $f(x, y)$  at  $\mathbf{a}$  is the *tangent plane*. If  $f_x(\mathbf{a}) = f_y(\mathbf{a}) = 0$  then the tangent plane is parallel to the  $xy$ -plane.
2. Suppose that  $\mathbf{a}$  is a saddle point of  $f(x, y)$ . Then there exists a curve

$$c : (-\epsilon, \epsilon) \rightarrow \mathbf{R}^2, \quad c(0) = (\mathbf{a})$$

such that composite function  $F(t) = (f \circ c)(t)$  has an inflection point at  $t = 0$ .

## Definition

Suppose that  $f(x, y)$  is of class  $\mathcal{C}^2$  at  $\mathbf{a} \in \mathbf{R}^2$ . Then

$$\Delta_f = f_{xx}(x_0, y_0)f_{yy}(x_0, y_0) - f_{xy}(x_0, y_0)^2$$

is called the **discriminant** of  $f$ .

## Example

Graph the following function at  $(0, 0)$  and compare their discriminants.

$$z = -x^2 - y^2, \quad z = x^2 + y^2, \quad z = x^2 - y^2$$

## Theorem (Hesse)

Let  $(x_0, y_0)$  be a critical point satisfying  $f_x(\mathbf{a}) = f_y(\mathbf{a}) = 0$ .

- ▶ If  $\Delta_f > 0$  and  $f_{xx}(x_0, y_0) > 0$ , then  $(x_0, y_0)$  is a local minimal point.
- ▶ If  $\Delta_f > 0$  and  $f_{xx}(x_0, y_0) < 0$ , then  $(x_0, y_0)$  is a local maximal point.
- ▶ If  $\Delta_f < 0$ , then  $f(x_0, y_0)$  is a saddle point.
- ▶ If  $\Delta_f = 0$ , then we cannot determine local extremity by this method.

### Remark

Let  $\Delta x = x - x_0$ ,  $\Delta y = y - y_0$ . The summands of the second degrees of  $Q_2(x, y)$  can be written as

$$\frac{1}{2} \begin{bmatrix} \Delta x & \Delta y \end{bmatrix} \underbrace{\begin{bmatrix} f_{xx}(x_0, y_0) & f_{xy}(x_0, y_0) \\ f_{xy}(x_0, y_0) & f_{yy}(x_0, y_0) \end{bmatrix}}_{=A} \begin{bmatrix} \Delta x \\ \Delta y \end{bmatrix}$$

Note that  $\Delta_f = \det A$ . Using linear transformation of  $x, y$ , we can transform the matrix  $A$  to one of three matrices below without changing the nature of extremities.

$$\begin{bmatrix} -1 & 0 \\ 0 & -1 \end{bmatrix}, \quad \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}, \quad \begin{bmatrix} \pm 1 & 0 \\ 0 & \mp 1 \end{bmatrix}$$

## Example (Least square method)

Suppose that we have a set of data

$$(x_1, y_1), (x_2, y_2), \dots, (x_n, y_n)$$

where each  $y_i$  is an experimental output for each input  $x_i$ .

Assume that hypothetically the output value  $y_j$  should follow a linear equation on the input value, say  $mx + b$ . Our goal is to find the value of  $m$  and  $b$  that best *fits* to the data: we want to minimize the sum of all squares of errors between the actual value  $y_i$  and true value  $mx_i + b$  to be minimum. In other words, we want

$$d(m, b) = \sum_{i=1}^n (y_i - (mx_i + b))^2$$

to be minimum.

Consider  $d(m, b)$  as two-variable function on  $m, b$ . The critical point is

$$m_0 = \frac{n \sum_{i=1}^n x_i y_i - \sum_{i=1}^n x_i \sum_{i=1}^n y_i}{n \sum_{i=1}^n x_i^2 - \left( \sum_{i=1}^n x_i \right)^2}, \quad b_0 = \frac{\sum_{i=1}^n x_i^2 \sum_{i=1}^n y_i - \sum_{i=1}^n x_i \sum_{i=1}^n x_i y_i}{n \sum_{i=1}^n x_i^2 - \left( \sum_{i=1}^n x_i \right)^2}$$

The Hessian of  $d(m, b)$  at  $(m_0, b_0)$  is

$$\begin{bmatrix} 2 \sum_{i=1}^n x_i^2 & 2 \sum_{i=1}^n x_i \\ 2 \sum_{i=1}^n x_i & 2n \end{bmatrix}$$

Since the discriminant of  $f$  is

$$\Delta_f(m_0, b_0) = 4n \sum_{i=1}^n x_i^2 - 4 \left( \sum_{i=1}^n x_i \right)^2 > 0$$

The point  $(m_0, b_0)$  is a local minimal point. In fact, it is a global minimal point. (why?)

## Theorem

Let  $(x_0, y_0, z_0)$  be a critical point of  $f(x, y, z)$  where  $f_x, f_y, f_z$  are all zero. Let  $H$  be the  $3 \times 3$  matrix defined by

$$\begin{bmatrix} f_{xx} & f_{xy} & f_{xz} \\ f_{yx} & f_{yy} & f_{yz} \\ f_{zx} & f_{zy} & f_{zz} \end{bmatrix}$$

Let  $d_1, d_2, d_3$  be the determinants of the  $1 \times 1, 2 \times 2, 3 \times 3$  sub-matrices on the left-top corner of  $H$ . Suppose that  $d_1 d_2 d_3 \neq 0$ .

- ▶ If  $d_i > 0$  for all  $i$ , then  $(x_0, y_0, z_0)$  is a local minimum point.
- ▶ if  $d_1, d_3 < 0$  and  $d_2 > 0$ , then  $(x_0, y_0, z_0)$  is a local maximal point.
- ▶ In all other cases,  $(x_0, y_0, z_0)$  is a saddle point.



## Proposition

*Let  $L_c(f)$  be the level curve at  $c = f(x_0, y_0)$ . on the  $xy$ -plane.  
Then the gradient vector  $\nabla f(x_0, y_0)$  is perpendicular to the  
curve  $L_c(f)$  at  $(x_0, y_0)$*

## Theorem (Lagrange multiplier)

*Let  $g(x, y)$ ,  $f(x, y)$  be differentiable functions. Let  $L_c(g)$  be a level curve at  $c$ . Let us restrict the domain of  $f$  onto  $L_c(g)$ . If  $(x_0, y_0)$  is an extremal point of  $f$  and  $\nabla g(x_0, y_0) \neq \mathbf{0}$ , there exists  $\lambda$  such that*

$$\nabla f(x_0, y_0) = \lambda \nabla g(x_0, y_0)$$

## Example

The Lagrange multiplier finds the maxima or minima of a **target** function  $f(x, y)$  under the **constraint**  $g(x, y) = c$ . Find the point on the circle  $x^2 + y^2 = 10$  where the function  $f(x, y) = 3x + y$  attains maximal or minimal.

## Corollary

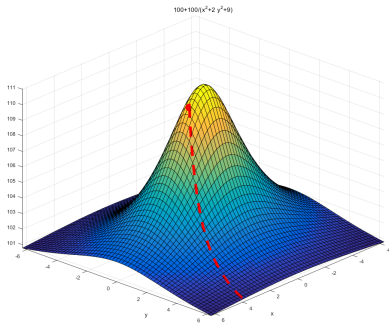
The direction of the gradient vector  $\nabla f(x_0, y_0)$  is the direction the value of function  $f(x, y)$  increases the fastest from the point  $(x_0, y_0)$ .

## Example

Let

$$f(x, y) = 100 + \frac{100}{x^2 + 2y^2 + 9}$$

be a function whose graph represent the contour of a mountain. Suppose that a water flow from the point  $(1, 0, 110)$  down the valley in the steepest direction. Find the trajectory of the water path.



## Theorem

*Let  $g(x, y, z)$ ,  $f(x, y, z)$  be differentiable functions. Suppose that  $(x_0, y_0, z_0)$  is a local extremal of  $f(x, y, z)$  restricted the level set  $L_c(g)$ . If  $\nabla g(x_0, y_0) \neq \mathbf{0}$ , then there exists  $\lambda$  such that*

$$\nabla f(x_0, y_0, z_0) = \lambda \nabla g(x_0, y_0, z_0)$$

## Example

Find the minimal and maximal value of  $f(x, y, z) = x^3 + y^3 + z^3$  on the sphere  $x^2 + y^2 + z^2 = 1$  on the first octant.

## Problem

Find all critical points and classify them

1.  $f(x, y) = xy + \frac{2}{x} + \frac{2}{y}$

2.  $e^y(x^2 + y^2 - z^2)$

## Problem

Find all local extremes of  $f(x, y)$  with the give constraints.

1.  $f(x, y) = 2x + y^2 + x^2, 2x^2 + y^2 = 2$
2.  $f(x, y, z) = xy + yz, x^2 + y^2 = 1, yz = 1$

## Problem

Find the local extremes of  $f(x, y) = x^2 + xy + y^2$  on the disk  $D = \{(x, y) \mid x^2 + y^2 \leq 1\}$ .



## Problem

Find the point on the graph  $xy^2z^3 = 2$  which is the closest to the origin.