

MAT1002 Lecture 17, Tuesday, Mar/26/2024

Outline

- Local extrema (14.7)
 - ↳ Definition
 - ↳ First-derivative test
 - ↳ Second-derivative test
- Global extrema of continuous functions (14.7)
- Optimization with equality constraints (14.8)
 - ↳ Lagrange multiplier

Optimization

Local Extrema

Definition

Let $f : D \rightarrow \mathbb{R}$, where $D \subseteq \mathbb{R}^n$, be a function, and let (a_1, \dots, a_n) be a point in D .

- We say that f has a **local maximum** at (a_1, \dots, a_n) if there exist an $r > 0$ such that

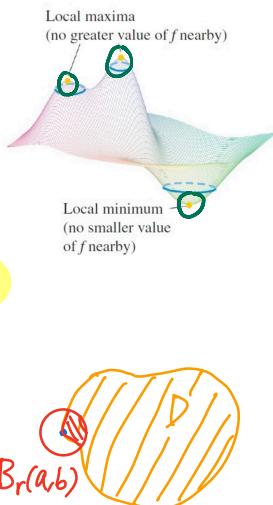
$$f(a_1, \dots, a_n) \geq f(x_1, \dots, x_n)$$

for every $(x_1, \dots, x_n) \in D \cap B_r(a_1, \dots, a_n)$. open ball with radius r

- We say that f has a **local minimum** at (a_1, \dots, a_n) if there exist an $r > 0$ such that

$$f(a_1, \dots, a_n) \leq f(x_1, \dots, x_n)$$

for every $(x_1, \dots, x_n) \in D \cap B_r(a_1, \dots, a_n)$.



Local extrema are also called relative extrema.

Caution: 1. A local extremum can occur at the interior or the boundary of D ; e.g., for $f(x, y) = x^2 + y^2$ defined on the closed disk

$$D = \{(x, y) : x^2 + y^2 \leq 1\},$$

the local maxima occur at the boundary of D : $\{(x, y) : x^2 + y^2 = 1\}$,

while the local minimum occur at $(x, y) = (0, 0)$, an interior point of D .

2. Behaviors for local extrema occurred at the boundary are different from those occurred at the interior, and we have different methods to find their candidates.

First Derivative Test - 阶导测试

Definition

An interior point (a, b) in the domain of a function f is called a **critical point** of f if either $\nabla f(a, b) = \vec{0}$ or at least one of $f_x(a, b)$ and $f_y(a, b)$ does not exist.

驻点一定是 \Rightarrow interior pt !!!! $\nabla f(a, b) = \vec{0}$

Although many theorems for **extrema** (i.e., maxima or minima) hold for n -variable functions, we will focus on these theorems for functions with 2 or 3 variables.

Theorem (First Derivative Test)

For a function $f : D \rightarrow \mathbb{R}$, where $D \subseteq \mathbb{R}^2$, suppose that:

- f has a local maximum or a local minimum at an interior point (a, b) of D , and;
- both $f_x(a, b)$ and $f_y(a, b)$ exist

必要条件 不是充分

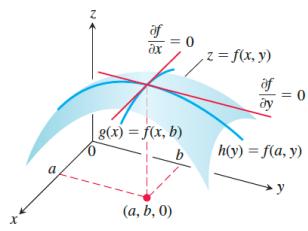
Then $f_x(a, b) = 0$ and $f_y(a, b) = 0$. That is, $\nabla f(a, b) = \vec{0}$.

In short, it states that "any local extremum of f that occurred at an interior point of D must occur at a critical point of f ".

Proof: • Define $g(x) := f(x, b)$. Then $g(a)$ is a local extremum, where a is an interior point of the domain of g . By one-variable theory, $g'(a) \text{ D.N.E.}$ or $g'(a) = 0$.

• $g'(a) = \frac{\partial}{\partial x} f(a, b)$, which exists by assumption, so $g'(a) = 0$
and
 $f_x(a, b) = 0$.

• Similarly, $f_y(a, b) = 0$.

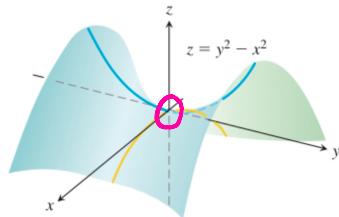


Note that the converse of the first-derivative test is not true:

a critical point of f does not always give a local extremum.

驻点 \nrightarrow 极值

e.g. For $f(x, y) = y^2 - x^2$, $(0, 0)$ is a critical point, but it does not give a local extremum. This is an example of a saddle point.



Def: Let $f: D \rightarrow \mathbb{R}$ ($D \subseteq \mathbb{R}^2$) be differentiable. If (a, b) is a critical point of f that does not give a local extremum of f , then (a, b) is called a **saddle point** of f , and $(a, b, f(a, b))$ is called a **saddle point** of the surface $z = f(x, y)$.

Smooth, CTS, Critical pt, no local max or no local min
twists in different direction

↙
 $\nabla \bar{f} = 0$

$\left\{ \begin{array}{l} f(x_0, y_0, \dots) \text{ not fixed} \\ \text{local extrema} \end{array} \right.$

不同方向看不一样

Second Derivative test

$$\text{Hess}(f) = \begin{bmatrix} f_{xx} & f_{xy} \\ f_{xy} & f_{yy} \end{bmatrix} \quad H(f) = \det[\text{Hess}(f)]$$

Theorem (Second Derivative Test)

Let f be a function whose second partial derivatives are all continuous on an open ~~ball~~ ^{disk} centered at (a, b) . Suppose that $\nabla f(a, b) = \vec{0}$ (so (a, b) is a critical point of f). Let

$$H := H(a, b) := f_{xx}(a, b)f_{yy}(a, b) - (f_{xy}(a, b))^2.$$

- ▶ If $H > 0$ and $f_{xx}(a, b) > 0$, then f has a local minimum at (a, b) .
- ▶ If $H > 0$ and $f_{xx}(a, b) < 0$, then f has a local maximum at (a, b) .
- ▶ If $H < 0$, then f has no local extremum at (a, b) ; that is, (a, b) is a saddle point of f .

of $H=0$ inconclusive.

local max/min

We postpone the proof to a later section. or saddle pt

Remark

If $H = 0$, the second derivative test gives no information: (a, b) could give a local maximum or a local minimum, or it could be a saddle point.

Example

$$f_x = 4x^3 - 4y = f_y = 4y^3 - 4x = 0.$$

Find all local extrema of the function

$$f(x, y) := x^4 + y^4 - 4xy + 1$$

defined on \mathbb{R}^2 .

$$f_{xx} = 12x^2 \quad f_{yy} = 12y^2$$

Sol: $(1, 1)$ and $(-1, -1)$, both give a local minimum.

$$f_{xy} = -4 \quad \text{local min}$$

$$144x^2y^2 - 16 > 0 \quad \text{Take } (0, 0)$$

$$f_{xx} > 0 \quad H(f) < 0 \quad \text{saddle pt}$$


 $f(x,y) = x - xy + y^2$ $x^2 + y^2 \geq 2xy$, $f(x,y) \geq 0$.
 $\begin{cases} \frac{\partial f}{\partial x} = 1 - y \\ \frac{\partial f}{\partial y} = 2y - x \end{cases} \Rightarrow y = x$ all pts are critical pts
 $f_{xx} = 0$ $f_{yy} = 2$ $A - B^2 = 0$ inconclusive
Global/Absolute Extrema $f_{xy} = -1$ so. all critical pts
 are local min.

Definition

Let (a_1, \dots, a_n) be a point in the domain D of a function f .

- We say that f has a **global maximum** (or an **absolute maximum**) at (a_1, \dots, a_n) if

$$f(a_1, \dots, a_n) \geq f(x_1, \dots, x_n)$$

for every $(x_1, \dots, x_n) \in D$.

- We say that f has a **global minimum** (or an **absolute minimum**) at (a_1, \dots, a_n) if

$$f(a_1, \dots, a_n) \leq f(x_1, \dots, x_n)$$

for every $(x_1, \dots, x_n) \in D$.

We saw in MAT1001 that a one-variable continuous function defined on a closed and bounded interval must attain its global extrema. The following is an analogous theorem for two-variable functions.

CTS + bdd + closed

Theorem

Let $f : D \rightarrow \mathbb{R}$ be a continuous function, where D is a closed and bounded set in \mathbb{R}^2 . Then f has a global maximum $f(x_1, y_1)$ and a global minimum $f(x_2, y_2)$ at some (x_1, y_1) and (x_2, y_2) in D .

In short, "continuous functions defined on closed and bounded sets attain their global extrema". Proof omitted.

Remark: D may not contain the absolute (global) extrema if it is not closed, even if it is bounded. e.g.

$$f(x,y) = x, \quad D = \{(x,y) : x^2 + y^2 < 1\}.$$

Since, by definition, a global maximum (or minimum) must be itself a local maximum (or minimum), the following is a strategy for finding the global extrema of a **continuous** function defined a closed and bounded domain D .

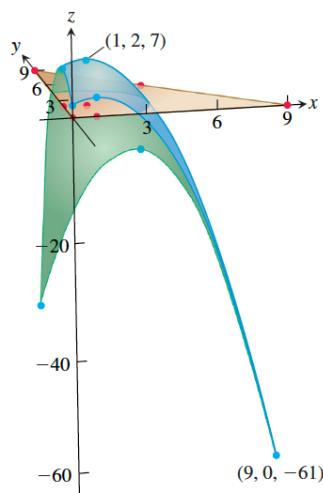
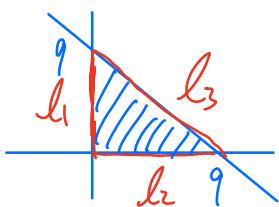
1. Find the values of f at the critical points (in the interior of the domain). *all candidates for local extrema in interior of D*
2. Find the local extrema of f on the boundary of D .
3. The largest value and the smallest value from step 1 and 2 are, respectively, the global maximum of f and the global minimum of f .

Example (e.g. 14.7.6)

Find the global maximum and global minimum of the function

$$f(x, y) := 2 + 2x + 4y - x^2 - y^2 \quad \leftarrow \text{cts}$$

defined on the triangular region in the first quadrant bounded by the lines $x = 0$, $y = 0$ and $y = 9 - x$.



Sol (Outline) :

• $f_x = 2 - 2x, f_y = 4 - 2y;$

$f_x = 0 = f_y \Leftrightarrow (x, y) = (1, 2)$; this is in interior of D.

• Boundary is $\ell_1 \cup \ell_2 \cup \ell_3$, where

$$\ell_1 := \{(0, y) : 0 \leq y \leq 9\}, \quad \ell_2 := \{(x, 0) : 0 \leq x \leq 9\},$$

$$\ell_3 := \{(x, 9-x) : 0 \leq x \leq 9\}.$$

• On ℓ_3 : $f(x, y) = f(x, 9-x) = -43 + 16x - 2x^2 = p(x);$

$$p'(x) = 0 \Leftrightarrow x=4 \Leftrightarrow y=5, \text{ so}$$

$(x, y) = (4, 5)$ is a candidate.

At end points $x=0$ and $x=9$, we have $(0, 9)$ and

$(9, 0)$ also being candidates. $f = 2 + 2x + 4y - x^2 - y^2$

• By doing similar arithmetic for ℓ_1 and ℓ_2 , $\ell_1 \rightarrow (0, y)$.

we found the following candidates:

$$\ell_1: (0, 0), (0, 2), (0, 9).$$

$$\ell_2: (0, 0), (1, 0), (9, 0).$$

$$f = 2 + 4y - y^2$$

$$g = 4 - 2y \quad y \neq 2$$

$$\ell_2 \rightarrow (x, 0) \quad (1, 0)$$

$$h = 2 + 2x - x^2$$

$$j = 2 - 2x \quad x \neq 1$$

• Compare all candidates:

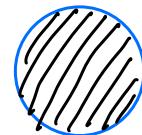
Max $\checkmark f(1, 2) = 7, f(0, 0) = 2, f(0, 2) = 6, f(0, 9) = -43,$

$$f(1, 0) = 3, f(9, 0) = -61, f(4, 5) = -11 \quad \checkmark$$

Optimization with Equality Constraints

Q: Given a mountain altitude function $f(x,y)$, what is the highest altitude given by points in the unit disk $x^2+y^2 \leq 1$? cts

- Find crit. pts. in the interior of D ,
- and find the maximum in the boundary,
- then compare.



$$D: x^2+y^2 \leq 1$$

Notation

$$\begin{array}{ll} \text{Maximize} & f(x,y) \\ \text{Subject to} & x^2+y^2=1. \end{array}$$

Boundary

This means we are trying to find $\max_{(x,y) \in C} f(x,y)$, where

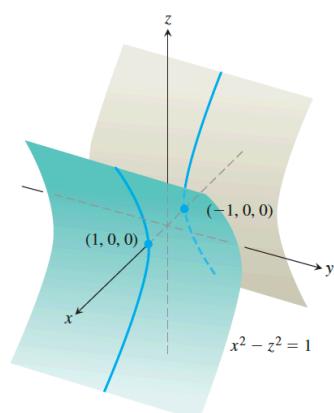
$$C = \{(x,y) : x^2+y^2=1\}.$$

Another example Find all points on the surface $x^2-z^2-1=0$ that are closest to the origin.

Notation

$$\text{Minimize } \sqrt{x^2+y^2+z^2}$$

$$\text{Subject to } x^2-z^2-1=0.$$



Optimization with One Equality Constraint

Maximize (or minimize) $f(x_1, \dots, x_n)$

Subject to $g(x_1, \dots, x_n) = 0$
(s.t.)

Consider Minimize $x^2 + y^2 + z^2$ s.t. $x^2 - z^2 - 1 = 0$. S

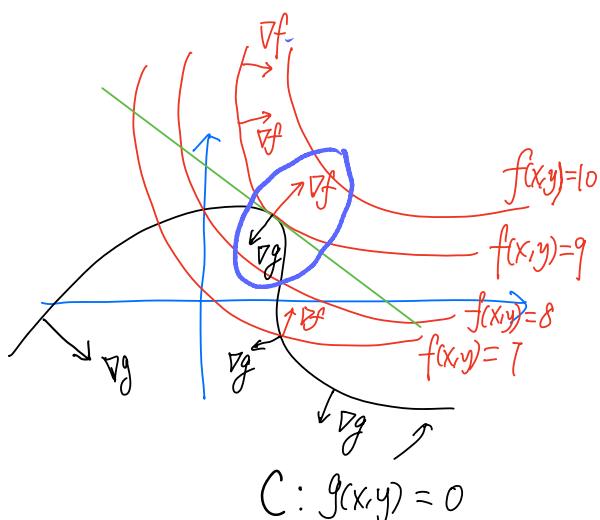
Critical pt method will not help, since for $f(x, y, z) = x^2 + y^2 + z^2$,
 $f_x = 2x$, $f_y = 2y$, $f_z = 2z$, so the only critical pt is
 $(x, y, z) = (0, 0, 0)$, which is not in S!

Need another method.

Method of Lagrange Multipliers

Consider a 2-variable case:

$$\begin{aligned} \text{Max } & f(x, y) \\ \text{s.t. } & g(x, y) = 0 \end{aligned}$$



Theorem (14.8.12, 2D Version)

$$\nabla g = 0 \text{ or } \nabla f = \lambda \nabla g$$

Let $f(x,y)$ be a differentiable function on D , and let C given by

$$(\text{Smooth}) \quad \vec{r}(t) = \langle x(t), y(t) \rangle, \quad t \in I$$

be a curve contained in the interior of D . If t_0 is an interior point of I and t_0 gives a local extremum of $f(\vec{r}(t)) = f(x(t), y(t))$, then at the point $P_0 := (x(t_0), y(t_0))$, we have $\nabla f \perp \vec{r}'$.

Proof: • Consider the values of f , but only on C .

• Let $g(t) := \underbrace{f(x(t), y(t))}_{\in C}$, and let $P_0 = (x(t_0), y(t_0))$.

• By assumption, at $t=t_0$, there is a local maximum of g . By MAT1001's one-variable optimization theory, we have $g'(t_0) = 0$.

• But $g'(t_0) = f_x(P_0)x'(t_0) + f_y(P_0)y'(t_0) = \nabla f(P_0) \cdot \vec{r}'(t_0)$ by the chain rule, so

$$\nabla f(P_0) \cdot \vec{r}'(t_0) = 0.$$

□

interior of C

In particular, when f has a local extremum at a point P_0

relative to C , where C is $g(x,y)=0$, then both ∇f and

∇g are \perp to the curve C at P_0 , So either :

- $\nabla g = \langle 0, 0 \rangle$, or ;
- $\nabla f = \lambda \nabla g$ for some $\lambda \in \mathbb{R}$ ($\nabla f \parallel \nabla g$). } at P_0

This observation suggests the following method.

The Method of Lagrange Multipliers

Suppose that the minimum of $f(x, y)$ subject to the constraint $g(x, y) = 0$ exists, where f and g are differentiable. Assume that $\nabla g(x, y) \neq \vec{0}$ whenever $g(x, y) = 0$. The following is a procedure for finding the constrained minimum:

$\min_{(x,y) \in C} f(x, y)$,
where $C := \{(x, y) : g(x, y) = 0\}$.

- Find all values of x, y and λ that simultaneously satisfy

$$\nabla f(x, y) = \lambda \nabla g(x, y) \quad \text{and} \quad g(x, y) = 0.$$

$$\nabla f = \lambda \nabla g$$

- Evaluate f at all the points (x, y) obtained in step 1.
The smallest value is the desired minimum.

If the constrained maximum exists, then the largest value in step 2 is the maximum.

$$y = \frac{\lambda}{4}x \quad y = \frac{\lambda^2}{4}y \\ x = \lambda y \quad y = 0 \\ \nabla f = \langle y, x \rangle \quad \text{or}$$

Example (e.g. 14.8.3) $xy + \lambda(\frac{x^2}{8} + \frac{y^2}{2} - 1) = 0$.

Find the maximum and minimum of the function

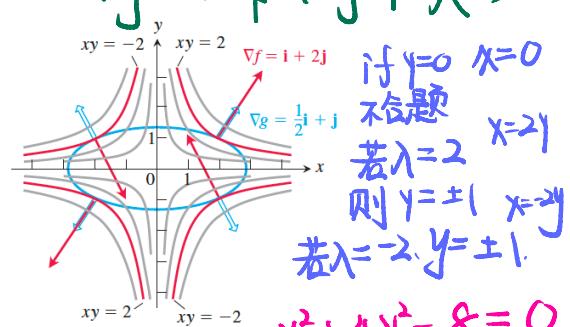
$$\begin{cases} fx = y + \frac{\lambda}{4}x = 0 \\ fy = x + \lambda y = 0 \\ f\lambda = \frac{x^2}{8} + \frac{y^2}{2} - 1 = 0 \end{cases}$$

defined on the ellipse $\frac{x^2}{8} + \frac{y^2}{2} = 1$.

Specify where the extrema occur on the ellipse.

Ans: Max at $(2, 1)$ and $(-2, -1)$; Min at $(2, -1)$ and $(-2, 1)$. $2x^2 = 8$

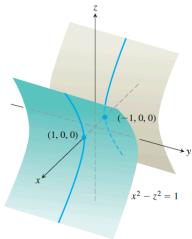
$$y = -\frac{\lambda}{4}x \quad x - \frac{\lambda}{4}x = 0 \quad \lambda = 2 \quad y = -\frac{1}{2}x \quad x = \pm 2 \\ x[1 - \frac{\lambda^2}{4}] = 0 \quad \text{or } -2 \quad y = \frac{1}{2}x \quad y = \pm 1 \\ \text{all pt } \max_{(2, 1)}, \min_{(2, -1)}, \min_{(-2, 1)}, \max_{(-2, -1)}$$



The method also works for functions with more variables.

比较法当且仅当 "candidate" 半所有 global max and global min

e.g. Find all points on the surface $S: x^2 - z^2 - 1 = 0$ that are closest to origin, and find the minimum distance.



Ans: $(\pm 1, 0, 0)$; distance = 1.

$$x^2 + y^2 + z^2 + \lambda(x^2 - z^2 - 1) = 0$$

$$fx = 2x + 2\lambda x = 0 \quad (\pm 1, 0, 0)$$

$$fy = 2y = 0 \rightarrow y = 0 \quad d = 1$$

$$fz = 2z - 2\lambda z = 0 \quad \begin{matrix} x^2 + y^2 + z^2 \\ \rightarrow \nabla f = \langle 2x, 2y, 2z \rangle \end{matrix}$$

$$f\lambda = x^2 - z^2 - 1 = 0 \quad x^2 - z^2 = 0$$

$$(x+1)(x-1) = z^2$$

$$(x+1) = z = 0$$

$$\text{or } (x-1) = z = 0$$

$$x = \pm 1$$

$$\begin{matrix} x^2 + y^2 + z^2 \\ \rightarrow \nabla g = \langle 2x, 2y, 2z \rangle \end{matrix}$$

$$\nabla f = \lambda \nabla g$$

$$\lambda = 1 \quad z = 0$$

$$x^2 - z^2 - 1 = 0$$

$$x = \pm 1$$

We need to say $r = \sqrt{z^2 + x^2 + y^2} = \sqrt{z^2 + 1 + y^2 + z^2}$

$$\sqrt{1 + 1} = 1$$

Hence we can say 1 is the shortest distance