

MAT1002 Lecture 18, Thursday, Mar/28/2024

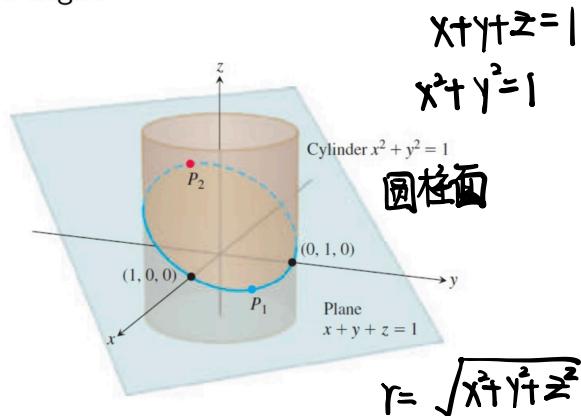
Outline

- Optimization with two equality constraints (14.8)
- Results related to second derivatives (14.9)
- Taylor's theorem for two-variable functions (14.9)

Two Equality Constraints 双等式限制

Example (e.g. 14.8.5)

The plane $x + y + z = 1$ cuts the cylinder $x^2 + y^2 = 1$ in an ellipse. Find the points on the ellipse that lie closest to and farthest from the origin.



That is : maximize/minimize $\sqrt{x^2 + y^2 + z^2}$ $x + y + z = 1$ $x^2 + y^2 = 1$
 Subject to $x + y + z = 1$ and $x^2 + y^2 = 1$.

This is an optimization problem with two equality constraints :

$$\begin{aligned} & \text{Max/min } f(x, y, z) \\ & \text{s.t. } g_1(x, y, z) = 0 \quad \text{and} \quad g_2(x, y, z) = 0 ; \end{aligned}$$

that is , finding extrema of f on the set $S_1 \cap S_2$, where

$$\underbrace{S_i = \{ (x, y, z) : g_i(x, y, z) = 0 \}}_{\text{a surface}}, \quad i \in \{1, 2\}.$$

If f has a local extremum at $P_0 \in S_1 \cap S_2$ relative to points in $S_1 \cup S_2$,

then $\nabla f(P_0)$ must be \perp to $C = S_1 \cap S_2$. Since ∇g_1 and

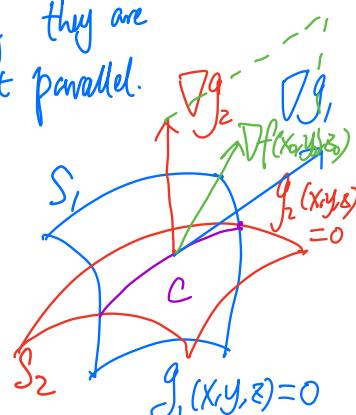
∇g_2 are also \perp to C , we have ∇f lying in the plane

spanned by $\nabla g_1(P_0)$ and $\nabla g_2(P_0)$.

Assuming they are
 $\neq 0$, not parallel.

This means that the method of Lagrange multipliers can still be used, but the system to be solved is the following instead:

$$\begin{cases} \nabla f(x, y, z) = \lambda_1 \nabla g_1(x, y, z) + \lambda_2 \nabla g_2(x, y, z) \\ g_1(x, y, z) = 0 \\ g_2(x, y, z) = 0 \end{cases}.$$



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

$$\text{即 } L(x, y, z, u, \lambda) = 0 \text{ 高数写法.}$$

$$\begin{cases} f_x = 0 & fu = 0 \\ f_y = 0 & fy = 0 \\ f_z = 0 & fz = 0 \\ x + y + z - 1 = 0 & f\lambda = 0 \end{cases}$$

Example (e.g. 14.8.5)

$$x + y + z = 1 \quad x^2 + y^2 = 1$$

The plane $x + y + z = 1$ cuts the cylinder $x^2 + y^2 = 1$ in an ellipse. Find the points on the ellipse that lie closest to and farthest from the origin.

$$\nabla f = \langle 2x, 2y, 2z \rangle$$

$$\nabla g_1 = \langle 1, 1, 1 \rangle$$

$$\nabla g_2 = \langle 2x, 2y, 0 \rangle$$

$$2x = \lambda + 2uz \Rightarrow 2z = \lambda + 2ux$$

$$2y = \lambda + 2uy \Rightarrow 2z = \lambda + 2uy$$

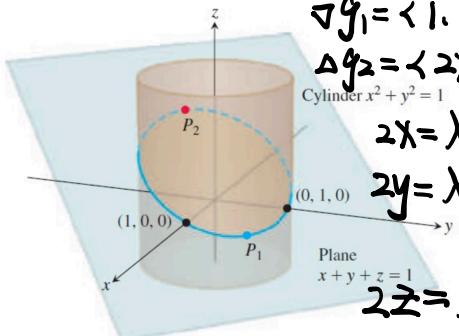
$$2z = \lambda + 0 \quad (1-u)x = z$$

$$(1-u)y = z$$

$$\begin{cases} x + y + z - 1 = 0 \\ x^2 + y^2 - 1 = 0 \end{cases}$$

$$\Rightarrow \begin{cases} x = 0 \text{ or } \\ y = 0 \end{cases}$$

这是个柱体!



Ans: Farthest: $(-\frac{\sqrt{2}}{2}, -\frac{\sqrt{2}}{2}, 1 + \sqrt{2})$.

$$u=1, z=0$$

Closest: $(0, 1, 0)$ and $(1, 0, 0)$

$$u=1 \quad x=y=\frac{z}{1-u}$$

$$x=y \quad 2x^2 - 1 = 0$$

$$x = \pm \frac{\sqrt{2}}{2}$$

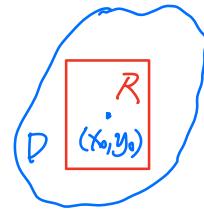
$$z = 1 \mp \sqrt{2}$$

Proof of Some Results Regarding Second-Order Partial

I. Error of Standard Linear Approximation 线性逼近

Result in short : $|E(x,y)| \leq \frac{1}{2}M((x-x_0)+(y-y_0))^2$, where M bounds the absolute values of all second partials on R .

(Assume all second partials are cts.) ($E(x,y) = f(x,y) - L(x,y)$)



Proof: • Let $h := \Delta x$ and $k := \Delta y$ be "small changes".

• Define $F(t) := f(\underbrace{x_0+th}_{x(t)}, \underbrace{y_0+tk}_{y(t)})$, $t \in [0,1]$.

• By Taylor's theorem, since f is cts on $[0,1]$ & diff'able on $(0,1)$,

$$\begin{aligned} \exists c \in (0,1) : F(1) &= F(0) + F'(0)(1-0) + \frac{1}{2}F''(c)(1-0)^2 \\ &= F(0) + F'(0) + \frac{1}{2}F''(c). \end{aligned} \quad (1)$$

$$\begin{aligned} F'(t) &= f_x x'(t) + f_y y'(t) \quad (\text{Chain rule, } F \text{ diff'able.}) \\ &= f_x h + f_y k. \end{aligned}$$

• Since all second partials of f are cts, F' is diff'able, so

$$\begin{aligned} F''(t) &= h \frac{d}{dt} f_x + k \frac{d}{dt} f_y \\ &= h(f_{xx} x'(t) + f_{xy} y'(t)) + k(f_{yx} x'(t) + f_{yy} y'(t)) \\ &= f_{xx} h^2 + 2f_{xy} hk + f_{yy} k^2 \end{aligned}$$

• Now ① \Rightarrow $L(x_0+h, y_0+k)$

$$f(x_0+h, y_0+k) = \underbrace{f(x_0, y_0) + f_x(x_0, y_0)h + f_y(x_0, y_0)k}_{\frac{1}{2}(f_{xx}h^2 + 2f_{xy}hk + f_{yy}k^2)} \Big|_{(x,y)=(x_0+h, y_0+k)} \\ E(x_0+h, y_0+k)$$

• In short notation, and by triangle inequality,

$$|E(x, y)| = \frac{1}{2} [f_{xx}(x-x_0)^2 + 2f_{xy}(x-x_0)(y-y_0) + f_{yy}(y-y_0)^2] \quad \begin{matrix} \text{Evaluated at} \\ (x_0+h, y_0+k) \\ \text{for some} \\ c \in (0,1) \end{matrix} \\ \leq \frac{1}{2} (f_{xx}|(x-x_0)^2| + 2|f_{xy}| |x-x_0| |y-y_0| + f_{yy} |(y-y_0)^2|) \\ \leq M (|x-x_0| + |y-y_0|)^2.$$

□

2. The Second Derivative Test 2阶导测试

Theorem (Second Derivative Test)

Let f be a function whose second partial derivatives are all continuous on an open ball 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 .

Proof of
this is
optional.

- Proof:
- Consider $h := \Delta x$ and $k := \Delta y$ as "small changes" from (a, b) .
 - From the first proof above, we saw that

$$\begin{aligned} f(a+h, b+k) &= f(a, b) + f_x(a, b)h + f_y(a, b)k \\ &\quad + \frac{1}{2}(f_{xx}h^2 + 2f_{xy}hk + f_{yy}k^2) \Big|_{(x,y)=(a+h, b+k)} \end{aligned}$$

for some $c \in (0, 1)$.

- Since $\nabla f(a, b) = (0, 0)$,

$$\begin{aligned} f(a+h, b+k) - f(a, b) &= \frac{1}{2}(f_{xx}h^2 + 2f_{xy}hk + f_{yy}k^2) \Big|_{(x,y)=(a+h, b+k)} \\ &=: Q(c, h, k) =: Q. \end{aligned} \tag{①}$$

Case 1: $H(a,b) = (f_{xx}f_{yy} - f_{xy}^2)(a,b) > 0$, $f_{xx}(a,b) > 0$.

- By continuity of second partials, $\exists \delta > 0$ s.t. $\forall (x_0, y_0) \in B_\delta(a, b)$, $H(x_0, y_0) > 0$ and $f_{xx}(x_0, y_0) > 0$.
- Consider (h, k) such that $0 < \sqrt{h^2 + k^2} < \delta$. Then

$$Q = \frac{1}{2} (f_{xx}h^2 + 2f_{xy}hk + f_{yy}k^2) \Big|_{(x,y)=(a+ch, b+ck)} \quad \begin{matrix} x_0 := \\ y_0 := \end{matrix}$$

$$\begin{aligned} \cdot f_{xx}(x_0, y_0) Q &= \frac{1}{2} (f_{xx}h^2 + 2f_{xx}f_{xy}hk + f_{xx}f_{yy}k^2) \Big|_{(x,y)=(x_0, y_0)} \\ &= \frac{1}{2} \left[(f_{xx}h + f_{xy}k)^2 + \underbrace{(f_{xx}f_{yy} - f_{xy}^2)k^2}_{H(x_0, y_0) > 0} \right] \Big|_{(x,y)=(x_0, y_0)} > 0. \end{aligned}$$

(since $(h, k) \neq (0, 0)$)

- Since $f_{xx}(x_0, y_0) > 0$, we have $Q > 0$. By ①,

$f(a+ch, b+ck) > f(a, b) \quad \forall (h, k) \text{ with } 0 < \sqrt{h^2 + k^2} < \delta$,
So (a, b) is a local minimum of f .

Case 2: $H(a,b) = (f_{xx}f_{yy} - f_{xy}^2)(a,b) > 0$, $f_{xx}(a,b) < 0$.

Similarly, we get $Q < 0$ and that (a, b) is a local maximum of f .

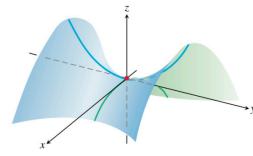
Case 3: $H(a,b) = (f_{xx}f_{yy} - f_{xy}^2)(a,b) < 0$

- We show that \exists direction \vec{w} and \vec{v} such that f has a local min at (a, b) while restricted to \vec{w} and has a local

max at (a,b) while restricted to \vec{v} . This would show that (a,b) is a saddle point of f .

- Let $\vec{u} := \langle h, k \rangle$ be a unit vector. Consider

$$F(t) := f(a+th, b+tk).$$



$$\text{Then } F'(0) = (f_x h + f_y k) \Big|_{(x,y)=(a,b)} = 0, \text{ and}$$

$$F''(0) = f_{xx} h^2 + 2f_{xy} hk + f_{yy} k^2 \Big|_{(x,y)=(a,b)}.$$

- Define $g(t) := \underbrace{f_{xx} t^2}_A + \underbrace{2f_{xy} t k}_B + \underbrace{f_{yy} k^2}_C$. Suppose $f_{xx}(a,b) \neq 0$.
 \leftarrow evaluated at (a,b) .

$$\text{Then } B^2 - 4AC = 4f_{xy}^2 - 4f_{xx}f_{yy} = -4(f_{xx}f_{yy} - f_{xy}^2) > 0.$$

This means that $g(t)$ has two distinct real roots, and $g(t)$ is sometimes > 0 and sometimes < 0 (depending on t). ②

- If $k \neq 0$, then $g\left(\frac{h}{k}\right) = f_{xx} \frac{h^2}{k^2} + 2f_{xy} \frac{h}{k} + f_{yy}$
 $\Rightarrow k^2 g\left(\frac{h}{k}\right) = f_{xx} h^2 + 2f_{xy} hk + f_{yy} k^2 = F''(0)$. ③
- By ② and ③, we may pick different combinations of (h,k) so that $F''(0) > 0$ on one and $F''(0) < 0$ on the other.
Concave up, local min in one direction Concave down, local max in another
- If $f_{xx}(a,b) = 0$, then $g(t) = 2f_{xy} t + f_{yy}$ with $f_{xy}(a,b) \neq 0$.
The rest is similar. Sometimes > 0 and Sometimes < 0 . □

泰勒定理.

Taylor's Theorem (Two-Variable Functions) with cts second-order partials

We saw that if we are given f and a fixed (a, b) , and let

$$F(t) := f(a+th, b+tk),$$

then

$$F'(t) = f_x h + f_y k = \left(h \frac{\partial}{\partial x} + k \frac{\partial}{\partial y} \right) f \Big|_{(a+th, b+tk)},$$

$$F''(t) = f_{xx} h^2 + 2f_{xy} hk + f_{yy} k^2 = \left(h \frac{\partial}{\partial x} + k \frac{\partial}{\partial y} \right)^2 f \Big|_{(a+th, b+tk)}.$$

More generally, if all n^{th} -order partials of f are cts on an open region containing (a, b) , then

$$F^{(n)}(t) = \left(h \frac{\partial}{\partial x} + k \frac{\partial}{\partial y} \right)^n f \Big|_{(a+th, b+tk)},$$

e.g. $F^{(4)}(t) = h^4 f_{xxxx} + 4h^3 k f_{xxyy} + 6h^2 k^2 f_{xyyy} + 4hk^3 f_{xyyy} + k^4 f_{yyyy}$. ↗

Evaluated at
 $(a+th, b+tk)$

One variable Taylor's theorem suggests that

$$F(1) = F(0) + F'(0) + \frac{1}{2!} F''(0) + \dots + \frac{1}{n!} F^{(n)}(0) + \frac{1}{(n+1)!} F^{(n+1)}(c)$$

for some $c \in (0, 1)$. This translates to the following two-variable

Version of Taylor's theorem: if $f(x, y)$ has continuous partials up to and including $(n+1)^{\text{th}}$ order in an open disk (or rectangular region) R containing (a, b) , then \forall points in R ,

$$f(a+th, b+tk) = \left(\sum_{i=0}^n \frac{1}{i!} \left(h \frac{\partial}{\partial x} + k \frac{\partial}{\partial y} \right)^i f \Big|_{(a,b)} \right) + \frac{1}{(n+1)!} \left(h \frac{\partial}{\partial x} + k \frac{\partial}{\partial y} \right)^{n+1} \Big|_{(a+th, b+tk)}$$

$$f(a+th, b+tk) = \left(\sum_{i=0}^n \frac{1}{i!} \left(h \frac{\partial}{\partial x} + k \frac{\partial}{\partial y} \right)^i f \Big|_{(a,b)} \right) + \frac{1}{(n+1)!} \left(h \frac{\partial}{\partial x} + k \frac{\partial}{\partial y} \right)^{n+1} \Big|_{(a+th, b+tk)}$$

for some $c \in (0, 1)$

$\text{EXAMPLE } 14.9.1$

Find a quadratic approximation to $f(x, y) = \sin x \sin y$ near the origin.
How accurate is the approximation if $|x| \leq 0.1$ and $|y| \leq 0.1$?

Sol.

$$f(0, 0) = \sin x \sin y|_{(0,0)} = 0, \quad f_{xx}(0, 0) = -\sin x \sin y|_{(0,0)} = 0,$$

$$f_x(0, 0) = \cos x \sin y|_{(0,0)} = 0, \quad f_{xy}(0, 0) = \cos x \cos y|_{(0,0)} = 1,$$

$$f_y(0, 0) = \sin x \cos y|_{(0,0)} = 0, \quad f_{yy}(0, 0) = -\sin x \sin y|_{(0,0)} = 0,$$

Approximation

- $f(x, y) \approx f(0, 0) + (xf_x + yf_y)|_{(0,0)} + \frac{1}{2}(x^2 f_{xx} + 2xy f_{xy} + y^2 f_{yy})|_{(0,0)}$
 $= xy$ $f_{xxx} = -\sin y \cos x$ $f_{xxy} = -\cos y \sin x$
- $E(x, y) = \frac{1}{6}(x^3 f_{xxx} + 3x^2 y f_{xxy} + 3xy^2 f_{xyy} + y^3 f_{yyy})|_{(cx, cy)}$
 $f_{xyy} = -\cos y \sin x$
- Computing all third-order partials shows that their absolute values are all ≤ 1 .
- Hence $|E(x, y)| \leq \frac{1}{6}((0.1)^3 + 3(0.1)^2 + 3(0.1)^3 + (0.1)^3) \leq 0.00134$.

Taylor's Formula for $f(x, y)$ at the Point (a, b)

Suppose $f(x, y)$ and its partial derivatives through order $n + 1$ are continuous throughout an open rectangular region R centered at a point (a, b) . Then, throughout R ,

$$\begin{aligned} f(a + h, b + k) &= f(a, b) + (hf_x + kf_y)|_{(a, b)} + \frac{1}{2!}(h^2 f_{xx} + 2hk f_{xy} + k^2 f_{yy})|_{(a, b)} \\ &\quad + \frac{1}{3!}(h^3 f_{xxx} + 3h^2 k f_{xxy} + 3hk^2 f_{xyy} + k^3 f_{yyy})|_{(a, b)} + \dots + \frac{1}{n!}\left(h \frac{\partial}{\partial x} + k \frac{\partial}{\partial y}\right)^n f|_{(a, b)} \\ &\quad + \frac{1}{(n+1)!}\left(h \frac{\partial}{\partial x} + k \frac{\partial}{\partial y}\right)^{n+1} f|_{(a+ch, b+ck)}. \end{aligned} \tag{7}$$

The first n derivative terms are evaluated at (a, b) . The last term is evaluated at some point $(a + ch, b + ck)$ on the line segment joining (a, b) and $(a + h, b + k)$.

If $(a, b) = (0, 0)$ and we treat h and k as independent variables (denoting them now by x and y), then Equation (7) assumes the following form.

Taylor's Formula for $f(x, y)$ at the Origin

$$\begin{aligned} f(x, y) &= f(0, 0) + xf_x + yf_y + \frac{1}{2!}(x^2 f_{xx} + 2xy f_{xy} + y^2 f_{yy}) \\ &\quad + \frac{1}{3!}(x^3 f_{xxx} + 3x^2 y f_{xxy} + 3xy^2 f_{xyy} + y^3 f_{yyy}) + \dots + \frac{1}{n!}\left(x^n \frac{\partial^n f}{\partial x^n} + nx^{n-1} y \frac{\partial^n f}{\partial x^{n-1} \partial y} + \dots + y^n \frac{\partial^n f}{\partial y^n}\right) \\ &\quad + \frac{1}{(n+1)!}\left(x^{n+1} \frac{\partial^{n+1} f}{\partial x^{n+1}} + (n+1)x^n y \frac{\partial^{n+1} f}{\partial x^n \partial y} + \dots + y^{n+1} \frac{\partial^{n+1} f}{\partial y^{n+1}}\right)|_{(cx, cy)} \end{aligned} \tag{8}$$