Introduction to Analysis Final preparation

0712238 林彥彤



回應: Final Exam 候選題

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- 21.7
- 21.8
- 24.1
- 24.5

HW12全

Theorem 21.7 (the high-dimensional binomial theorem). (exercise)

For two vectors $x = (x_1, \dots, x_k), y = (y_1, \dots, y_k) \in \mathbb{R}^k$ and a multi-index $\gamma = (\gamma_1, \dots, \gamma_k) \in \mathbb{N}_0^k$, we have $(x + y)^{\gamma} = \sum_{\alpha, \beta \in \mathbb{N}_0^k, \alpha + \beta = \gamma} \frac{\gamma!}{\alpha! \beta!} x^{\alpha} y^{\beta} = \sum_{\alpha_i + \beta_i = \gamma_i \text{ for all } 1 \le i \le k} \frac{\gamma_1! \dots \gamma_k!}{\alpha_1! \dots \alpha_k! \beta_1! \dots \beta_k!} x_1^{\alpha_1} \dots x_k^{\alpha_k} y_1^{\beta_1} \dots y_k^{\beta_k}$.

$$\begin{aligned} & \textbf{Proposition 21.8} \text{ (the product rule for partial derivatives). } (\textit{exercise}) \\ & \textit{Let } S \subset \mathbb{R}^k \text{ be open, } f,g:S \to \mathbb{R} \text{ be } C^r \text{ differentiable on } S, \text{ and } \gamma = (\gamma_1,\cdots,\gamma_k) \in \mathbb{N}_0^k. \\ & \textit{Then } \partial^\gamma (fg) = \sum_{\alpha,\beta \in \mathbb{N}_0^k,\alpha+\beta=\gamma} \frac{\gamma!}{\alpha!\beta!} (\partial^\alpha f) (\partial^\beta g) = \sum_{\substack{\alpha_i+\beta_i=\gamma_i \text{ for all } 1 \leq i \leq k \\ 0 \leq \alpha_i,\beta_i \leq \gamma_i}} \frac{\gamma_1! \cdots \gamma_k!}{\alpha_1! \cdots \alpha_k!\beta_1! \cdots \beta_k!} (\partial_1^{\alpha_1} \partial_2^{\alpha_2} \cdots \partial_k^{\alpha_k} f) (\partial_1^{\beta_1} \partial_2^{\beta_2} \cdots \partial_k^{\beta_k} g). \end{aligned}$$

Theorem 24.1 (the uniqueness of the Taylor polynomial).

Let $S \subset \mathbb{R}^k$ be an open convex set, $c = (c_1, c_2, \dots, c_k) \in S$, and $m \ge 1$ be an integer.

Let $f: S \to \mathbb{R}$ be C^m on S and f(x) = Q(x-c) + E(x-c) for all $x \in S$,

where $Q : \mathbb{R}^k \to \mathbb{R}$ is a polynomial in x - c of degree $\leq m$

where $Q : \mathbb{R}^k \to \mathbb{R}$ is a posynomial in \mathbb{R} and $E : \mathbb{R}^k \to \mathbb{R}$ is a function with $\lim_{x \to c} \frac{E(x - c)}{\|x - c\|^m} = 0$.

Then for any
$$x = (x_1, x_2, \dots, x_k) \in S$$
,

$$Q(x-c) = \sum_{\substack{n=0 \\ \alpha_1+\alpha_2+\dots+\alpha_k=n \\ \alpha_1 \in \alpha}}^{m} \frac{\partial_1^{\alpha_1} \partial_2^{\alpha_2} \cdots \partial_k^{\alpha_k} f(c)}{\alpha_1! \alpha_2! \cdots \alpha_k!} (x_1-c_1)^{\alpha_1} (x_2-c_2)^{\alpha_2} \cdots (x_k-c_k)^{\alpha_k},$$

i.e., Q(x - c) is the mth-order Taylor polynomial for f about c on S.

If k = 1, we may only require f is C^{m-1} on S and $f^{(m)}(c)$ exists. (exercise)

Theorem 24.5 (extreme value theorem for C^r functions). (exercise) Let $f:(a,b) \to \mathbb{R}$ be a C^r function on (a,b), where $r \in \mathbb{N}$, and let $c \in (a,b)$. If $f^{(n)}(c) = 0$ for all $1 \le n < r$ and $f^{(r)}(c) \ne 0$, then the followings hold.

- 1. If r is even and $f^{(r)}(c) > 0$, then f has a local minimum at c.
- If r is even and f^(r)(c) < 0, then f has a local maximum at c.
- 3. If r is odd, then f has neither a local minimum nor a local maximum at c.

Exercise 12

2019/12/24

- Find the maximum and minimum values of f(x, y) = x² + x + 2y² on the
 unit circle.
- Let A ∈ R^{m×n} with m ≥ n and rank(A) = n, b ∈ R^m. Solve the following optimization problem: x* = arg min_x ||Ax b||².
- Let A ∈ R^{m×n} with m ≤ n and rank(A) = m, b ∈ R^m. Solve the following optimization problem: x* = arg min_x ||x||² subject to Ax = b
- Suppose f: R² → R² is defined as f(u,v) = (u³ + uv + v³, u² v²). Show that there is a neighborhood U of (u, v) = (1, 1) on which f|_U: U → f(U) is diffeomorphic. Calculate D((f|_U)⁻¹).
- Theorem 26.7

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Theorem 26.7. (C^1 \text{ functions})

Let S \subset \mathbb{R}^k be open and f: S \to \mathbb{R}^m with f = (f_1, \dots, f_m).

Then the partial derivatives \partial_j f_i all exist and are continuous on S

\iff f is continuously differentiable on S,

that is, f is differentiable on S and Df: S \to L(\mathbb{R}^k, \mathbb{R}^m) is continuous on S

(in this case we say that f is C^1 on S or f \in C^1(S)). (advanced exercise)
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https://www.youtube.com/watch?time_continue=247&v=et2sTg9qslw&feature=emb_logo CORRESPONDING TO CHAPTER 26

Pf. Let
$$\eta(x) = \begin{cases} \frac{f(x) - f(a) - A(x-a)}{|x-a|_{A_{1}}} & \text{if } x \neq a \\ 0 & \text{if } x = a. \end{cases}$$

Then $f(x) = f(a) + A(x-a) + \eta(a) |x-a|_{d_{2}}$

all x .

$$f(a_{1}, a_{1} + h_{1}, a_{n}) \quad \mathbb{R}^{m}$$

Ans: No in general.

Consider
$$\mathbb{R}^2$$
 f \mathbb{R}

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

$$\frac{\partial f}{\partial x}(0,0) = 0 = \frac{\partial f}{\partial y}(0,0), \quad \text{of } f(x,y) = (0,0), \quad \text{hence not at } (0,0), \quad \text{of } f(x,y) = (0,0$$

Theorem (the chain rule) Given $S \xrightarrow{f} T \xrightarrow{g} R^l$ and $A \in M_{m,n}(IR)$, $B \in M_{d,m}(R)$, if $(f(x) \sim f(a) + A(x-a) \text{ as } x \to a \implies g(f(x)) \sim g(f(x)) + BA(x-a)$ $(g(y) \sim g(f(a)) + B(y-f(a)) \text{ as } y \to f(a) \text{ as } x \to a.$ - More explicitly, 2 8 (f(x, x,n), ..., fm(x, xn)) (x,... $= \sum_{s=1}^{m} \frac{\partial g}{\partial y} (f_{i}(a_{s}, a_{n}), \dots, f_{n}(a_{s}, a_{n})) \frac{\partial f_{s}}{\partial x} (a_{s}, \dots, a_{n})$

Ex.
$$f$$
 is continuously diff on S and $a \in S$ \Longrightarrow f is diff. at a .

Example Consider
$$f(x) = \chi^{\frac{3}{2}}$$
, $\chi > 0$.

$$f(\chi) \cdot g(\chi, \chi_2) = \chi^{\frac{1}{2}} (\chi > 0)$$

$$\frac{\partial g}{\partial \chi} = \chi_2 \chi^{\frac{1}{2}}, \quad \frac{\partial g}{\partial \chi} = \chi^{\frac{1}{2}} \ln \chi \implies g \cdot \text{contily.}$$

$$f(\chi) = g(\chi, \chi^3) \quad f(\chi) = \frac{\partial g}{\partial \chi} (\chi, \chi^3) (\frac{1}{\chi}) \frac{\partial g}{\partial \chi} (\chi, \chi^3) (\chi, \chi^3)$$

Example Let
$$A(x) = e^{x^5 e^x}$$

Compute $T_{o,13}(x) = T_{o,13}^f(x)$

Let $f(x) = x^5 e^x = x^5 \left(1 + x + \frac{x^3}{2!} + \dots + \frac{x^8}{8!}\right)$
 $f(x) = e^x$
 $f(x) = e^x$
 $f(x) = e^x$
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 $f(x) = e^x$

To simplify life, use the $\|\cdot\|_1$ norm. To reduce clutter, let $\phi_k(h) = (h_1, \dots, h_k, 0, \dots 0)$. Let $\phi_0(h) = 0$, and note that $\|\phi_k(h) - \phi_j(h)\|_1 \le \|h\|_1$.

First suppose m = 1.

Let $\epsilon > 0$.

W By continuity, we can choose $\delta>0$ such that $|D_kf(a+h)-D_kf(a)|<\epsilon$ for all k and $\|h\|<\delta$.

Let $A = (D_1 f(a), \ldots, D_n f(a))$ and suppose $\|h\| < \delta$, then we have

$$egin{aligned} |f(a+h)-f(a)-Ah| &= |\sum_{k=1}^n (f(a+\phi_k(h))-f(a+\phi_{k-1}(h))-D_kf(a)h_k)| \ &\leq \sum_{k=1}^n |f(a+\phi_k(h))-f(a+\phi_{k-1}(h))-D_kf(a)h_k| \end{aligned}$$

By the mean value theorem, there are $c_k \in [a+\phi_{k-1}(h), a+\phi_k(h)]$ (that is, each c_k lies on the line segment) such that $f(a+\phi_k(h)) - f(a+\phi_{k-1}(h)) = D_k f(c_k) h_k$. Note that $\|c_k - a\|_1 \le \|h\|_1$. Continuing:

$$|f(a+h) - f(a) - Ah| \le \sum_{k=1}^{n} |D_k f(c_k) h_k - D_k f(a) h_k|$$

$$= \sum_{k=1}^{n} |D_k f(c_k) - D_k f(a)| |h_k|$$

$$< \epsilon \sum_{k=1}^{n} |h_k|$$

$$= \epsilon ||h||_1$$

Since $\epsilon > 0$ was arbitrary, this shows that f is differentiable at a and Df(a)h = Ah.

It is straightforward to show that if $f_1, \ldots f_m$ are differentiable, then so is $f(x) = (f_1(x), \ldots, f_n(x))$, and $Df(x)h = (Df_1(x)h, \ldots, Df_m(x)h)$.

differentiable but not continuous differentiable A good example of such a function is $x_2(\sin(1/x_2))$ which has a finite derivative at x=0,

but the derivative is essentially discontinuous at x=0

Taylor expansion with integral remainder
$$f \in C^n(I)$$

open interval— I f is n -times continuous differentiable on I

(i.e. $f, f, ..., f^n$ all exist and are $f(b) = f(a) + \int_a^b \frac{1}{(b-b)} f(b) dt$
 $f(a) = f(a) + (t-b) f(b) \int_{t=a}^{b-b} \frac{1}{a} f(b) dt$

https://www.voutube.com/watch?time_continue=1&v=dvFziACt_iE&feature=emb_logo

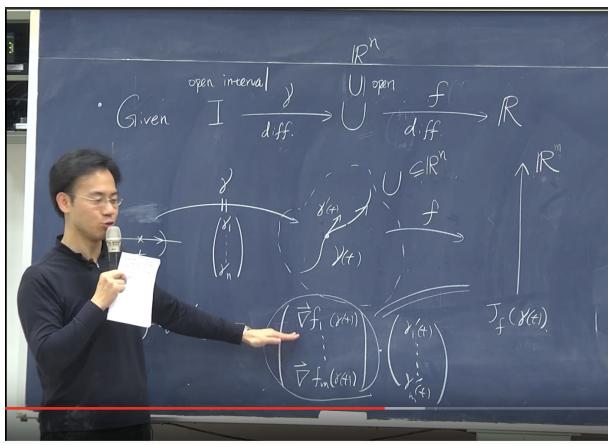
積分餘項

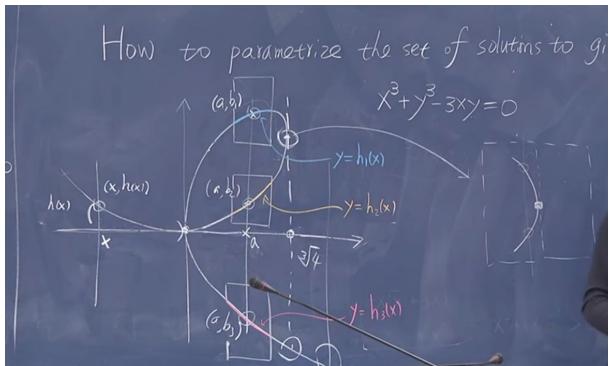
多變數泰勒

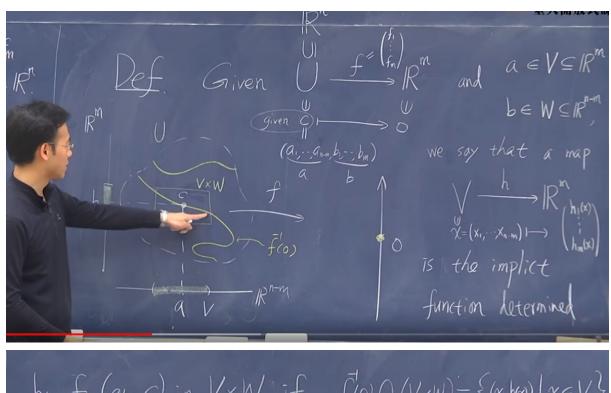
【微積分一/齊震宇老師2015/12/22A】Taylor展開式餘項的積分表達

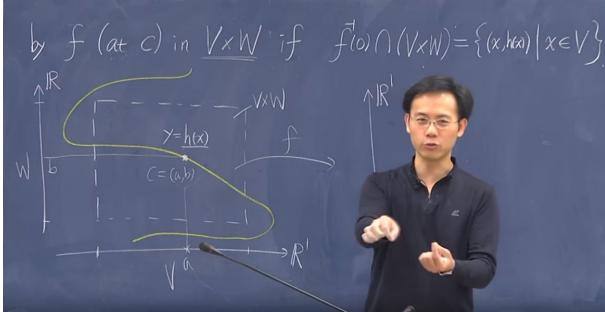
https://www.voutube.com/watch?v=O4n9laJ2mR8

微積分二:【多變數微分理論2】梯度向量;(積分與偏導數混合版本的)均值定理









https://www.youtube.com/watch?time_continue=2&v=LF7hXMB5I78&feature=emb_logo

Implicit function theosem (With
$$m=2$$
, $n=3$)
$$(x,y,z) R^{3} = (f_{1},f_{2})$$

$$(a,b,c) \longrightarrow (0,0)$$

$$\exists y \in \mathbb{R}$$

教授您好:

注意到課程網站說1/9號13:00會送學期成績至註冊組但1/10期末考卷才能確認是否應該在課堂確認考卷沒問題之後再送成績比較好由於期末考很多科忘記提早寄信很抱歉

彥彤 敬上