

矩阵函数

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定义 $f: M_n(\mathbb{C}) \rightarrow M_n(\mathbb{C})$

$$A \rightarrow f(A)$$

(1) 多项式函数

$$f(z) = a_0 + a_1 z + \dots + a_n z^n$$

$$\downarrow$$

$$f(A) = a_0 I + a_1 A + \dots + a_n A^n$$

性质 (1) 若 $AP=B$, 则 $P^{-1}AP = f(B)$

若 $P^{-1}AP = J$, 则 $J = P^{-1}AP$

$$\Rightarrow f(A) = P \left(\begin{smallmatrix} f(J) \\ & f(J) \\ & & f(J) \\ & & & f(J) \end{smallmatrix} \right) P^{-1}$$

定理: 设 $\alpha_0 + \alpha_1 z + \dots + \alpha_n z^n$, $J = \begin{pmatrix} \alpha_0 & & & \\ & \alpha_1 & & \\ & & \ddots & \\ & & & \alpha_n \end{pmatrix}$

$$P^{-1}f(A)P = \begin{pmatrix} \alpha_0 I & & & \\ & \alpha_1 A & & \\ & & \ddots & \\ & & & \alpha_n A^n \end{pmatrix}$$

证明: $J = \begin{pmatrix} \alpha_0 & & & \\ & \alpha_1 & & \\ & & \ddots & \\ & & & \alpha_n \end{pmatrix}$

(2) 一般矩阵函数定义

$$f(x) = e^x \rightarrow e^A$$

幂级数收敛半径 R

定理: 存在 $0 < R \leq \infty$, 使得对于 $0 < |z| < R$, 级数收敛, 对于 $|z| \geq R$, 级数发散

Taylor 定理:

$$f(z) = f(0) + f'(0) + \frac{1}{2!} f''(0) + \frac{1}{3!} f'''(0) + \dots + \frac{1}{k!} f^k(0)$$

$$\text{例 } e^z = \sum \frac{z^n}{n!} (n \geq 0)$$

$$\frac{1}{1-z} = \sum z^n (n \geq 0)$$

$$\sin z = \sum \frac{(-1)^n z^n}{(2n+1)!} (n \geq 0)$$

$$\cos z = \sum \frac{(-1)^n z^n}{(2n)!} (n \geq 0)$$

$$\ln(1+z) = \sum \frac{(-1)^n z^n}{n+1} (n \geq 0)$$

$$\sin z = \frac{e^{iz} - e^{-iz}}{2} \quad \cos z = \frac{e^{iz} + e^{-iz}}{2}$$

$S(z) \rightarrow S(A)$?

$$S_A = a_0 + a_1 A + a_2 A^2 + \dots + a_n A^n$$

$$S_A = a_0 I + a_1 A + a_2 A^2 + \dots + a_n A^n$$

性质 (1) 若 $AP=B$, 则

$S(A) \text{ 收敛} \Leftrightarrow S(B) \text{ 收敛}$.

(2) $S(A)$ 收敛 $\Leftrightarrow S(A) = S(\lambda_i)$ 收敛

(3) $S(A)$ 收敛半径, $J = \begin{pmatrix} \alpha_0 & & & \\ & \alpha_1 & & \\ & & \ddots & \\ & & & \alpha_n \end{pmatrix}$

当 $\lambda_i < r$, $S(\lambda_i)$ 收敛.

$$S(A) = \begin{pmatrix} S(\lambda_1) & & & \\ & S(\lambda_2) & & \\ & & \ddots & \\ & & & S(\lambda_n) \end{pmatrix}$$

定义: $A \in M_n(\mathbb{C})$

$$m_A(x) = (x - \lambda_1)^{m_1} \cdots (x - \lambda_n)^{m_n}$$

$$\left\{ \begin{array}{l} S(\lambda_1) = \cdots = S^{(m_1)} \\ S(\lambda_2) = \cdots = S^{(m_2)} \\ \vdots \\ S(\lambda_n) = \cdots = S^{(m_n)} \end{array} \right.$$

存在 λ_i 称 $S(\lambda_i)$ 为 A 的单根

例 $A = \begin{pmatrix} 3 & 0 & -1 \\ 1 & 1 & 1 \\ 0 & 0 & 1 \end{pmatrix}$ 求 e^A , $\sin A$.

解: $|A - \lambda I - A| = (1 - \lambda)^3$.

$$P = \begin{pmatrix} 0 & 1 & 0 \\ 1 & -1 & 1 \\ 0 & 1 & -1 \end{pmatrix} \quad P^{-1} = \begin{pmatrix} 0 & 1 & 1 \\ 1 & 0 & 0 \\ 0 & -1 & 1 \end{pmatrix}$$

$$P^{-1}AP = \begin{pmatrix} 2 & 1 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix} = J.$$

$$S(A) = P S(J) P^{-1} \quad S(J) = S_0 \quad S_0 = \begin{pmatrix} S(2) & S(1) \\ 0 & S(1) \end{pmatrix}$$

定理: 设 $A \in M_n(\mathbb{C})$, $\deg m_A(x) = m \geq 2$,

若 e^A 为 A 的单根 $\leq m-1$ 的多项式,

即 $e^A = I + A + \frac{A^2}{2!} + \dots + \frac{A^m}{m!} -$

$$m_A(x) = x^m + a_{m-1} x^{m-1} + \dots + a_1 x + a_0$$

$$\Rightarrow A^m = -[a_{m-1} A^{m-1} + a_{m-2} A^{m-2} + \dots + a_1 A + a_0 I]$$

$$A^{m-1} = -$$

即 A 的 m 阶特征值可用 $\leq m-1$ 的多项式表示.

(待证)

性质: (1) A 为支点 $\Rightarrow e^A \cdot e^B = e^{A+B}$

(2) e^A 可逆, 逆为 e^{-A} .

(3) A 为 Hermitian, 则 e^A 为 skew-Hermitian.

$\Rightarrow e^A$ 为反演.

(类似 $b \rightarrow i\theta \rightarrow e^{i\theta}$)

- 阶常系数微分方程组

$$\text{设 } A \in M_n(\mathbb{C}) \quad x(t) = \begin{pmatrix} x_1(t) \\ x_2(t) \\ \vdots \\ x_n(t) \end{pmatrix}$$

$$\text{解得, } \frac{dx(t)}{dt} = Ax(t)$$

$$\text{其中 } \frac{dx(t)}{dt} = \begin{pmatrix} x_1'(t) \\ x_2'(t) \\ \vdots \\ x_n'(t) \end{pmatrix}$$

$$\text{例, } A = \begin{pmatrix} 1 & 2 \\ 3 & 4 \end{pmatrix}$$

$$\text{解得: } \begin{pmatrix} \frac{dx_1(t)}{dt} \\ x_1'(t) \end{pmatrix} = Ax(t)$$

$\Leftrightarrow x_1'(t) = \mathbb{C}^n$ (齐次)

同解方程且唯一

$$x_1(t) = e^{At} x_1(0) \Leftrightarrow x_1(t) = (e^{At}) x_1(0)$$

$$e^{At} = P e^{\lambda t} P^{-1} = P \begin{pmatrix} e^{\lambda_1 t} & & \\ & \ddots & \\ & & e^{\lambda_n t} \end{pmatrix} P^{-1}$$

$$\lambda_1 = 1, \lambda_2 = 2$$

$$P = \begin{pmatrix} 1 & 1 \\ 1 & -1 \end{pmatrix}$$

$$P^{-1} = \begin{pmatrix} 1 & 0 \\ 0 & 2 \end{pmatrix}$$

$$e^{\lambda_1 t} = e^t$$

$$e^{\lambda_2 t} = e^{2t}$$

$$e^{At} = P e^{\lambda t} P^{-1} = P \begin{pmatrix} e^t & & \\ & \ddots & \\ & & e^{2t} \end{pmatrix} P^{-1}$$

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