

(1)

Problem 1

a)

(1)

$$u''' = t^2 u u'' - u v'$$

$$v''' = t v v' + 4 u'$$

set:

$$y_1 = u$$

$$y_2 = u'$$

$$y_3 = u''$$

$$y_4 = v$$

$$y_5 = v'$$

(2)

$$\underline{y}' = f(t, \underline{y}) \Leftrightarrow \begin{pmatrix} y_1 \\ y_2 \\ y_3 \\ y_4 \\ y_5 \end{pmatrix}' = \begin{pmatrix} y_2 \\ y_3 \\ t^2 y_1 y_3 - y_1 y_5 \\ y_5 \\ t y_4 y_5 + 4 y_2 \end{pmatrix}$$

b)

$$\underline{f}_{\underline{y}} = \begin{pmatrix} \frac{\partial f_1}{\partial y_1} & \dots & \frac{\partial f_1}{\partial y_5} \\ \vdots & & \vdots \\ \frac{\partial f_5}{\partial y_1} & \dots & \frac{\partial f_5}{\partial y_5} \end{pmatrix} = \begin{pmatrix} 0 & 1 & 0 & 0 & 0 \\ 0 & 0 & 1 & 0 & 0 \\ t^2 y_3 - y_5 & 0 & t y_1 & 0 & -y_1 \\ 0 & 0 & 0 & 0 & 1 \\ 0 & 4 & 0 & t y_5 & t y_4 \end{pmatrix}$$

c) f Lipschitz iff $\|f(t, x) - f(t, y)\| \leq \lambda \|x - y\| \quad \forall x, y \in \mathbb{R}^d, t \geq t_0$

If $f, x, y \in \mathbb{R}$ the mean value theorem gives:

$$f(t, x) - f(t, y) = \frac{\partial f}{\partial x}(t, \xi) (x - y)$$

Thus if $|\frac{\partial f}{\partial x}| \leq \lambda \quad \forall x, y \in \mathbb{R}$ we have $|f(t, x) - f(t, y)| \leq \lambda |x - y|$

and the function is Lipschitz. The mean value theorem does not extend straightforward to $x, y \in \mathbb{R}^d$ in higher dimensions.

But let us do a multi-dim Taylor expansion:

$$f(t, y) = f(t, x + (y - x)) = f(t, x) + \underline{f}_y(\xi) (y - x)$$

$$\Rightarrow f(t, x) - f(t, y) = -\underline{f}_y(\xi) (y - x) \quad \|\cdot\|$$

(2)

$$\|f(t, x) - f(t, y)\| = \|f_y(\xi) \cdot (x - y)\| \leq \|f_y(\xi)\| \|x - y\|$$

Thus if we can bound $\|f_y(\xi)\|$ for any ξ we have a Lipschitz constant, here ≤ 1

$\|y\|_1 \leq 1 \Rightarrow |y_i| \leq 1$, Thus all components in f_y is bounded and so is $\|f_y\|$. The smallest possible λ depends on the norm we choose.

Problem 2

a) $y' = Ay$; $A \in \mathbb{R}^{s \times s}$ & $A = A^T \Rightarrow \exists Q$ orthogonal ($Q^T Q = I$)
 $y(0) = y_0$ $A = Q^T D Q$, set $u = Qy =$

Exact solution: $Qy' = QAy$
 $u' = Du \Rightarrow u(t) = \begin{pmatrix} e^{\lambda_1 t} & & 0 \\ & \ddots & \\ 0 & & e^{\lambda_s t} \end{pmatrix} u_0 = E(t) y_0$

Euler:

$$y_{n+1} = y_n + hAy_n = (I + hA)y_n \Rightarrow y_n = (I + hA)^n y_0 = [Q^T (I + hD) Q]^n y_0$$

Error:

$$\begin{aligned} e_n &= y_n - y(t_n) = \left((I + hA)^n - E(t) \right) y_0 \\ \|e_n\|_2 &\leq \| (I + hA)^n - E(t) \|_2 \|y_0\|_2 \\ &= \| Q^T [(I + hD)^n - E(t)] Q \|_2 \|y_0\|_2 \\ &= \| (I + hD)^n - E(t) \|_2 \|y_0\| \\ &= \max_{\lambda \in \sigma(A)} |(1 + h\lambda)^n - e^{nh\lambda}| \|y_0\| \end{aligned}$$

Note 1:

$$\begin{aligned} (I + hA)^n &= Q^T (I + hD)^n Q \\ \text{then } (I + hA)^n &= Q^T (I + hD) Q Q^T (I + hD) Q \dots Q^T (I + hD) Q \\ &= Q^T (I + hD)^n Q \end{aligned}$$

Note 2: Since $E(t)$ is a diagonal matrix

$$E(t) = Q^T E(t) Q$$

Note 3: $\|Q^T B Q\|_2 = \|B\|$ whenever Q is orthogonal

Note 4: $\|D\|_2$; D diagonal $= \max |d_i|$

Remark: Contrary to the estimate in proof of Th. 1.1 this is a good estimate for the error.

(3)

(b)

$$e^x = 1 + x + \frac{x^2}{2}; \quad x \in [-1, 0] \Rightarrow e^x \geq (1+x) \Rightarrow e^{nx} \geq (1+x)^n$$

$$e^x = 1 + x + \frac{x^2}{2} + \frac{x^3}{3!}; \quad x \in [-1, 0] \Rightarrow e^x \leq 1 + x + \frac{x^2}{2}$$

proves the
right inequality

same
even

$$(a-b)^n = a^n - n a^{n-1} b + \binom{n}{2} a^{n-2} b^2 - \binom{n}{3} a^{n-3} b^3 + \dots$$

$$= a^n - n a^{n-1} b + \sum_{k=1}^{\frac{n}{2}} a^{n-2k-1} b^k \left(\binom{n}{2k} a - \binom{n}{2k+1} b \right)$$

$$||| \quad |a-1| \ll 1, |b| \ll 1$$

$$\Rightarrow a \gg |b|$$

$$\text{and } \left(\binom{n}{2k} a - \binom{n}{2k+1} b \right) > 0$$

$$\geq a^n - n a^{n-1} b$$

$$\text{Set } a = 1 + x + \frac{x^2}{2}; \quad b = \frac{x^2}{2} \quad \text{then:}$$

$$\begin{aligned} (1+x)^n &= (a-b)^n \geq a^n - n a^{n-1} b = \left(1+x+\frac{x^2}{2}\right)^n - n \left(1+x+\frac{x^2}{2}\right)^{n-1} \frac{x^2}{2} \\ &\geq e^{xn} - \frac{nx^2}{2} e^{x(n-1)} \quad (**) \end{aligned}$$

c) From a.

$$\|e_n\|_2 \leq |(1+h\lambda_{\max})^n - e^{nh\lambda_{\max}}| \|y_0\|$$

$$nh = t$$

$$\leq |e^{nh\lambda_{\max}} - \frac{n}{2} (h\lambda_{\max})^2 e^{nh\lambda_{\max}(n-1)} - \cancel{nh\lambda_{\max}}| \|y_0\|$$

$$= \frac{1}{2} |t \lambda_{\max}^2 \cdot e^{(t-h)\lambda_{\max}}| h \|y_0\|$$

$$\leq \frac{1}{2} t^* h \lambda_{\max}^2 \|y_0\| \quad (\text{since } \lambda_{\max} < 0 \quad e^{t\lambda_{\max}} < 1)$$

d) eigen values of $A = \begin{bmatrix} -2 & 1 \\ 1 & -2 \end{bmatrix}$ are $(-3, -1)$

From formula p. 7 (here the λ is the Lipschitz constant

which will be equal to $\|A\|_2 = |\lambda_{\max}| = 3$

$$\|e_{n,h}\|_2 \leq \frac{C}{3} (e^{3t^*} - 1) h; \quad \text{From above}$$

$$\|e_n\|_2 \leq \frac{1}{2} t^* h 3^2 \|y_0\|$$

linear
vs. exponential
(t^*)