

Thm If $f(t, x)$ is Lipschitz in x and the exact sol'n $x(t) \in C^2$,

then for fixed T , the numerical sol'n x_n by forward Euler w/

$$h = \frac{T - t_0}{n} \text{ satisfies}$$

$$|x(T) - x_n| \leq C(T) \cdot h$$

Proof Denote $e_i = x(t_i) - x_i$

$$x_{i+1} = x_i + h f(t_i, x_i)$$

depending on $\|x(t)\|_{C^2([t_0, T])}$

$$x(t_{i+1}) = x(t_i) + h f(t_i, x(t_i)) + O(h^2)$$

$$\Rightarrow e_{i+1} = e_i + h \underbrace{\left(f(t_i, x(t_i)) - f(t_i, x_i) \right)}_{\substack{\uparrow \\ \text{Lip. const. of } f}} + O(h^2)$$

$$|\cdot| \leq L \cdot |x(t_i) - x_i| = L \cdot |e_i|$$

Lip. const. of f

$$|e_{i+1}| \leq |e_i| + hL \cdot |e_i| + Ch^2 \quad e_0 =$$

$$= (1 + Lh) |e_i| + Ch^2 \quad \text{depending on } \|x(t)\|_{C^2([t_0, T])}$$

$$|e_n| \leq (1 + Lh) |e_{n-1}| + Ch^2$$

$$\leq (1 + Lh) \left((1 + Lh) |e_{n-2}| + Ch^2 \right) + Ch^2$$

$$\leq \dots \leq Ch^2 \left(1 + (1 + Lh) + (1 + Lh)^2 + \dots + (1 + Lh)^{n-1} \right)$$

$$= Ch^2 \frac{(1 + Lh)^n - 1}{(1 + Lh) - 1}$$

$$= h \cdot \frac{C}{L} \left((1 + Lh)^{\frac{1}{Lh} \cdot Lhn} - 1 \right)$$

$$\leq h \cdot \frac{C}{L} \left(e^{Lhn} - 1 \right)$$

$$\left(1 + \alpha \right)^{1/\alpha} < e \quad \forall \alpha > 0$$

$$\leq h \cdot \underbrace{\frac{C}{L} \left(e^{L(T-t_0)} - 1 \right)}_{C(T)}$$

exponential growth in T

"Gronwall inequality"

$$x' \leq A x + \underline{\quad}$$

$$x(t) \leq \underline{\quad}$$

8.2 Taylor series method

$$x'(t) = f(t, x(t))$$

$$x(t+h) = x(t) + h \underbrace{x'(t)}_{f(t, x(t))} + \frac{h^2}{2} x''(t) + \frac{h^3}{6} x'''(t) + \dots$$

$f(t, x(t))$

forward
Euler
L.T.E. = $O(h^2)$

2nd order
Taylor
L.T.E. = $O(h^3)$

$$x''(t) = \partial_t f(t, x(t)) + \partial_x f(t, x(t)) \cdot x'(t)$$

$$= \partial_t f(t, x(t)) + \partial_x f(t, x(t)) \cdot f(t, x(t))$$

2nd order Taylor method:

$$x_{i+1} = x_i + h f(t_i, x_i) + \frac{h^2}{2} \left[\partial_t f(t_i, x_i) + \partial_x f(t_i, x_i) \cdot f(t_i, x_i) \right]$$

$$x'''(t) = \partial_{tt} f(t, x(t)) + \partial_{tx} f(t, x(t)) \cdot x'(t)$$

$$+ f(t, x(t)) \cdot \left(\partial_{tx} f(t, x(t)) + \partial_{xx} f(t, x(t)) \cdot x'(t) \right)$$

$$+ \partial_x f(t, x(t)) \cdot \left(\partial_t f(t, x(t)) + \partial_x f(t, x(t)) \cdot x'(t) \right)$$

→ 3rd order Taylor method.

- Taylor methods are only applicable when partial derivatives of f are easily calculated (be careful about vector-valued ODEs!)

8.3 Runge-Kutta method

Forward Euler $x^* = x_i + h f(t_i, x_i)$

Try: $x_{i+1} = x_i + b_1 h f(t_i, x_i) + b_2 h f(t_i+h, x^*)$

Choose coeffs b_1, b_2 so that Taylor expansion of x_{i+1} and $x(t_{i+1})$ match up to $O(h^2)$ (L.T.E. = $O(h^2)$, 2nd order accuracy)

$$x(t_{i+1}) = x(t_i) + h x'(t_i) + \frac{h^2}{2} x''(t_i) + \dots$$

$$= \boxed{x(t_i)} + \boxed{h f(t_i, x(t_i))} + \boxed{\frac{h^2}{2} \left(\partial_t f(t_i, x(t_i)) + \partial_x f(t_i, x(t_i)) \cdot f(t_i, x(t_i)) \right)} + \dots$$

$$x_{i+1} = x_i + b_1 h f(t_i, x_i) + b_2 h f(t_i+h, x_i + h f(t_i, x_i))$$

$$= x_i + \underline{b_1 h f(t_i, x_i)} + b_2 h \left(\underline{f(t_i, x_i)} + \underline{\partial_t f(t_i, x_i) \cdot h} \right. \\ \left. + \underline{\partial_x f(t_i, x_i) \cdot h f(t_i, x_i)} + O(h^2) \right)$$

$$= \boxed{x_i} + \boxed{h f(t_i, x_i) \cdot (b_1 + b_2)} \\ + \boxed{h^2 \cdot b_2 \cdot \left(\partial_t f(t_i, x_i) + \partial_x f(t_i, x_i) \cdot f(t_i, x_i) \right)} + O(h^3)$$

$$\begin{cases} b_1 + b_2 = 1 \\ b_2 = \frac{1}{2} \end{cases} \quad b_1 = \frac{1}{2}$$

$$\Rightarrow \text{Heun's method (second order)} \quad \begin{cases} x^* = x_i + h f(t_i, x_i) \\ x_{i+1} = x_i + \frac{1}{2} h f(t_i, x_i) + \frac{1}{2} h f(t_i+h, x^*) \end{cases}$$

General framework of Runge-Kutta

$$x^{(1)} = x_i$$

$$x^{(2)} = x_i + a_{21} h f(t_i, x^{(1)}) \quad c_2 := a_{21}$$

$$x^{(3)} = x_i + a_{31} h f(t_i, x^{(1)}) + a_{32} h f(t_i + c_2 h, x^{(2)}) \quad c_3 := a_{31} + a_{32}$$

$$x^{(4)} = x_i + a_{41} h f(t_i, x^{(1)}) + a_{42} h f(t_i + c_2 h, x^{(2)}) + a_{43} h f(t_i + c_3 h, x^{(3)})$$

...

$$\left\{ \begin{array}{l} x^{(j)} = x_i + h \sum_{k=1}^{j-1} \underline{a_{jk}} f(t_i + c_k h, x^{(k)}) \\ x_{i+1} = x_i + h \sum_{k=1}^s \underline{b_k} f(t_i + c_k h, x^{(k)}) \end{array} \right. \quad \begin{array}{l} c_j = \sum_{k=1}^{j-1} a_{jk} \\ j = 1, \dots, s \end{array}$$

Butcher table :

0	0	0	...	0	0
c_2	a_{21}	0	...	0	0
c_3	a_{31}	a_{32}	...	0	0
\vdots	\vdots				\vdots
c_s	a_{s1}	a_{s2}	...	$a_{s,s-1}$	0
	b_1	b_2	...	b_{s-1}	b_s

Heun's :

0	0	0
1	1	0
	$\frac{1}{2}$	$\frac{1}{2}$