

The concept of stability and convergence can be understood by analysing the numerical solutions of linear equation $u' = \lambda u$ corresponding to non-linear equation $u' = f(\lambda u)$. This linear form $u' = \lambda u$ is called test function.

Consider IVP (linear

$$u' = \lambda u, \quad u(t_0) = \eta_0 \quad \text{--- ①}$$

exact solution is $u(t) = \eta_0 e^{\lambda t}$

$$u(t_{j+1}) = \eta_0 e^{\lambda t_{j+1}}$$

$$u(t_j) = \eta_0 e^{\lambda t_j}$$

$$\frac{u(t_{j+1})}{u(t_j)} = e^{\lambda(t_{j+1} - t_j)} = e^{\lambda h}$$

$$\text{or } u(t_{j+1}) = e^{\lambda h} u(t_j) \quad \text{--- ③}$$

⑦ A numerical method can be obtained by writing an approximation to $e^{\lambda t}$ say

$$(i) \quad e^{\lambda h} \approx 1 + \lambda h$$

$$(ii) \quad e^{\lambda h} \approx 1 + \lambda h + \frac{\lambda^2 h^2}{2}$$

$$(iii) \quad e^{\lambda h} \approx 1 + \lambda h + \frac{\lambda^2 h^2}{2} + \frac{\lambda^3 h^3}{6}$$

$$(iv) \quad e^{\lambda h} \approx 1 + \lambda h + \frac{\lambda^2 h^2}{2} + \frac{\lambda^3 h^3}{6} + \frac{\lambda^4 h^4}{24}$$

$$(v) \quad e^{\lambda h} \approx \frac{1 + \lambda h/2}{1 - \lambda h/2}$$

$$\left. \begin{aligned} & \text{④} \\ & + \frac{\lambda^4 h^4}{24} \end{aligned} \right\}$$

Consider the problem $u' = \lambda u$ i.e., $f(x, u) = \lambda u$

(1*)

Euler method

$$u_{j+1} = u_j + hf = u_j + h\lambda u_j = (1 + \lambda h)u_j$$

Taylor's second order method

$$\begin{aligned} u_{j+1} &= u_j + h u_j' + \frac{h^2}{2!} u_j'' \\ &= u_j + h\lambda u_j + \frac{h^2}{2!} \lambda^2 u_j \\ &= \left[1 + \lambda h + \frac{(\lambda h)^2}{2!} \right] u_j \end{aligned}$$

$$\begin{aligned} u' &= f = \lambda u \\ u'' &= \frac{d}{dx} u' = \lambda u' \\ &= \lambda f \\ &= \lambda \cdot \lambda u \\ &= \lambda^2 u \\ u''' &= \lambda^2 u' = \lambda \cdot \lambda u \\ &= \lambda^3 u \end{aligned}$$

Similarly for higher order methods.

So a numerical method can be obtained by writing an approximation to $e^{\lambda h}$ say

(i) $e^{\lambda h} \approx 1 + \lambda h$

Euler method

(ii) $e^{\lambda h} \approx 1 + \lambda h + \frac{\lambda^2 h^2}{2!}$

Taylor's second order method

Stability

Let these approximations for $e^{\lambda h}$ be denoted by $E(\lambda h)$ then the methods will be written as

$$u_{j+1} = E(\lambda h) u_j \quad \text{--- (5)}$$

and the exact solution will satisfy

$$u(t_{j+1}) = E(\lambda h) u(t_j) + T_j \quad \text{--- (6)}$$

Let $\epsilon_j = u_j - u(t_j)$ then $u_j = \epsilon_j + u(t_j)$ and from (5) we get

$$\begin{aligned} \epsilon_{j+1} + u(t_{j+1}) &= E(\lambda h) (\epsilon_j + u(t_j)) \\ \epsilon_{j+1} &= -u(t_{j+1}) + E(\lambda h) \epsilon_j + E(\lambda h) u(t_j) \end{aligned}$$

from (3) $u(t_{j+1}) = e^{\lambda h} u(t_j)$

$$\begin{aligned} \text{Then } \epsilon_{j+1} &= -e^{\lambda h} u(t_j) + E(\lambda h) u(t_j) + E(\lambda h) \epsilon_j \\ &= \underbrace{[E(\lambda h) - e^{\lambda h}] u(t_j)}_{\text{Truncation error}} + \underbrace{E(\lambda h) \epsilon_j}_{\text{Propagation of the error}} \end{aligned}$$

Next for $\epsilon_{j+2} = [E(\lambda h) - e^{\lambda h}] u(t_{j+1}) + E(\lambda h) \epsilon_{j+1}$
Now $u(t_{j+1}) = e^{\lambda h} u(t_j)$ so

$$\epsilon_{j+2} = (E(\lambda h) - e^{\lambda h}) e^{\lambda h} u(t_j) + E(\lambda h) [E(\lambda h) - e^{\lambda h}] u(t_j) + E^2(\lambda h) \epsilon_j$$

$$\boxed{\epsilon_{j+2} = \underbrace{[E^2(\lambda h) - e^{2\lambda h}] u(t_j)}_{\text{Truncation error}} + \underbrace{E^2(\lambda h) \epsilon_j}_{\text{Propagation of error}}}$$

for first term on r.h.s.

$$u(t_{j+1}) = e^{\lambda h} u(t_j)$$

$$u(t_{j+2}) = e^{\lambda h} u(t_{j+1}) = e^{2\lambda h} u(t_j)$$

and numerical sol.

$$u_{j+1} = E(\lambda h) u_j$$

$$u_{j+2} = E(\lambda h) u_{j+1} = E^2(\lambda h) u_j$$

$$\text{Thus Truncation error} = [E^2(\lambda h) - e^{2\lambda h}] u(t_j)$$

Now at $j+k$ step.

$$E_{j+k} = \underbrace{(E^k(\lambda h) - e^{k\lambda h}) u(t_j)}_{\text{Truncation error}} + \underbrace{E^k(\lambda h) E_j}_{\text{Propagation of the error.}}$$

Defⁿ Stable A method is stable if cumulative effect of all errors, including round off errors is bounded, independent of mesh divisions.

Absolutely stable If $|E(\lambda h)| < 1$ then method is

called absolutely stable.

Relatively stable If $E(\lambda h) < e^{\lambda h}$ then the method is called relatively stable.


Now (i) If $E(\lambda h) = 1 + \lambda h < e^{\lambda h}$ for $\lambda > 0$
 and hence for $\lambda > 0$ is relatively stable (always).
 for Absolute Stability

$$|E(\lambda h)| < 1$$

$$|1 + \lambda h| < 1$$

$$\text{or } -1 < 1 + \lambda h < 1$$

$$\text{or } -2 < \lambda h < 0$$

Thus for $\lambda h \in (-2, 0)$ the method is absolutely stable and this interval is called interval of stability. 

$$(ii) E(\lambda h) = 1 + \lambda h + \frac{\lambda^2 h^2}{2}$$

Again for $\lambda > 0$ $E(\lambda h) < e^{\lambda h}$ hence
 for $\lambda > 0$ the method is relatively stable
 for Absolute Stability

$$|E(\lambda h)| < 1$$

$$\left| 1 + \lambda h + \frac{\lambda^2 h^2}{2} \right| < 1$$

$$\text{or } \left| \frac{1}{2} [(1 + \lambda h)^2 + 1] \right| < 1$$

$$\begin{aligned} & 1 + \lambda h + \frac{\lambda^2 h^2}{2} \\ & \frac{1}{2} (\lambda^2 h^2 + 2\lambda h + 2) \\ & = \frac{1}{2} [(1 + \lambda h)^2 + 1] \end{aligned}$$

$$\text{or } -1 < \frac{1}{2} [(1 + \lambda h)^2 + 1] < 1$$

$$-2 < (1 + \lambda h)^2 + 1 < 2$$

The left hand side inequality is always satisfied for R.H.S.

Stability

$$(1+\lambda h)^2 + 1 < 2$$

$$\text{or } (1+\lambda h)^2 < 1$$

$$-1 < (1+\lambda h) < 1$$

$$-2 < \lambda h < 0$$

So interval of stability is $(-2, 0)$.

(iii) Similarly for

$$E(\lambda) = 1 + \lambda h + \frac{\lambda^2 h^2}{2} + \frac{\lambda^3 h^3}{6}$$

$$\lambda h = -2.5$$

$$\frac{6.25}{2}$$

$$\left(1 - 2.5 + \frac{3.125}{2} - 2.6042 \right)$$

$$= -0.979$$

| | | | | |
|----------------|---|----|----|---------------|
| λh | 0 | -1 | -2 | -2.5 |
| $E(\lambda h)$ | 1 | | | <u>-0.979</u> |

So $\lambda h \in (-2.5, 0)$.

(iv) Similarly for

$$E(\lambda h) = 1 + \lambda h + \frac{\lambda^2 h^2}{2} + \frac{\lambda^3 h^3}{6} + \frac{\lambda^4 h^4}{24}$$

| | | | | | | |
|----------------|---|--------|-------|--------|--------|-------|
| λh | 0 | -1 | -2 | -2.2 | -2.6 | -3.0 |
| $E(\lambda h)$ | 1 | 0.3750 | 0.333 | 0.4212 | 0.7547 | 1.375 |

(v) $E(\lambda) = \frac{1 + \lambda h_2}{1 - \lambda h_2}$

we want

$$|E(\lambda)| < 1 \quad \text{or} \quad \left| \frac{1 + \lambda h_2}{1 - \lambda h_2} \right| < 1$$

$$\text{or } |1 + \lambda h_2| < |1 - \lambda h_2| \quad \text{--- (*)}$$

for $\lambda < 0$ inequality (*) is always satisfied. So interval of absolute stability $(-\infty, 0)$. Such method

are called. Stability
unconditionally stable.

(6)

Euler Method

$y_{n+1} = y_n + h f_n$
 Apply Euler method for $y' = \lambda y$

$$y_{n+1} = y_n + \lambda h y_n$$

$$y_{n+1} = (1 + \lambda h) y_n$$

$$E(\lambda h) = 1 + \lambda h$$

2nd Order R-K method

$$y_{n+1} = y_n + \frac{1}{4} (k_1 + 3k_2)$$

$$k_1 = h f(x_n, y_n)$$

$$k_2 = h f(x_n + \frac{3}{2}h, y_n + \frac{2}{3}k_1)$$

Apply R-K method for $y' = \lambda y$

$$k_1 = h \lambda y_n$$

$$k_2 = h \lambda (y_n + \frac{2}{3}k_1)$$

$$= h \lambda (y_n + \frac{2}{3} h \lambda y_n)$$

$$= [h \lambda + \frac{2}{3} (h \lambda)^2] y_n$$

$$y_{n+1} = y_n + \left[\frac{1}{4} h \lambda + \frac{3}{4} (h \lambda) + \frac{\frac{3}{4}}{2} \cdot \frac{2}{3} (h \lambda)^2 \right] y_n$$

$$= \left[1 + h \lambda + \frac{h^2 \lambda^2}{2} \right] y_n$$

$$E(\lambda h) = 1 + h \lambda + \frac{h^2 \lambda^2}{2}$$

4th order R-K method

$$y_{n+1} = y_n + \frac{1}{6} [k_1 + 2k_2 + 2k_3 + k_4]$$

$$k_1 = hf(x_n, y_n)$$

$$k_2 = hf(x_n + \frac{1}{2}h, y_n + \frac{1}{2}k_1)$$

$$k_3 = hf(x_n + \frac{1}{2}h, y_n + \frac{1}{2}k_2)$$

$$k_4 = hf(x_n + h, y_n + k_3)$$

Apply this method to $y' = \lambda y$

$$k_1 = h\lambda y_n$$

$$\begin{aligned} k_2 &= hf(x_n + \frac{1}{2}h, y_n + \frac{1}{2}k_1) \\ &= h\lambda [y_n + \frac{1}{2}h\lambda y_n] \\ &= [h\lambda + \frac{1}{2}(h\lambda)^2] y_n \end{aligned}$$

$$\begin{aligned} k_3 &= h\lambda (y_n + \frac{1}{2}k_2) \\ &= h\lambda [y_n + \frac{1}{2}(h\lambda + \frac{1}{2}(h\lambda)^2) y_n] \\ &= [h\lambda + \frac{1}{2}(h\lambda)^2 + \frac{1}{4}(h\lambda)^3] y_n \end{aligned}$$

$$\begin{aligned} k_4 &= h\lambda (y_n + k_3) \\ &= h\lambda [y_n + (h\lambda + \frac{1}{2}(h\lambda)^2 + \frac{1}{4}(h\lambda)^3) y_n] \\ &= [h\lambda + (h\lambda)^2 + \frac{1}{2}(h\lambda)^3 + \frac{1}{4}(h\lambda)^4] y_n \end{aligned}$$

$$\begin{aligned} y_{n+1} &= y_n + \frac{1}{6} (k_1 + 2k_2 + 2k_3 + k_4) \\ &= y_n + \frac{1}{6} [h\lambda + 2\{h\lambda + \frac{1}{2}(h\lambda)^2\} + 2\{h\lambda + \frac{1}{2}(h\lambda)^2 + \frac{1}{4}(h\lambda)^3\} \\ &\quad + \{h\lambda + (h\lambda)^2 + \frac{1}{2}(h\lambda)^3 + \frac{1}{4}(h\lambda)^4\}] \end{aligned}$$

$$y_{n+1} = \left[1 + h\lambda + \frac{1}{6}(h\lambda)^2 (1+1+1) + \frac{1}{6}(h\lambda)^3 \left(\frac{1}{2} + \frac{1}{2}\right) + \frac{1}{6 \cdot 4}(h\lambda)^4 \right] y_n \quad (3)$$

$$y_{n+1} = \left[1 + h\lambda + \frac{(h\lambda)^2}{2} + \frac{(h\lambda)^3}{6} + \frac{(h\lambda)^4}{24} \right] y_n$$

$$E(\lambda) = 1 + h\lambda + \frac{(h\lambda)^2}{2} + \frac{(h\lambda)^3}{6} + \frac{(h\lambda)^4}{24}$$