

Suppl. lecture. (should watch this lecture by Next Tuesday)

HW2

Ch. 4 Numerical solution of ordinary differential eqs. (ODEs)

$$y'' + \omega^2 y = f(x)$$

$$\left\{ \begin{array}{l} y(0) = y_0 \\ \left. \frac{dy}{dx} \right|_0 = v \end{array} \right. \quad \text{or}$$

↑  
initial value problem

(all the conditions are given at one point.)

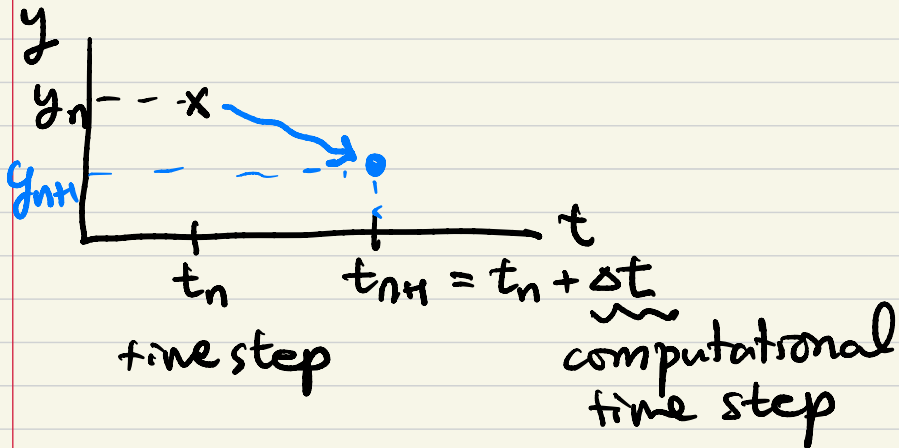
$$\left\{ \begin{array}{l} y(0) = y_0 \\ y(L) = y_L \end{array} \right.$$

↑  
boundary value problem

(conditions are given at more than one pt.)

## 4.1 Initial value problems

$$\begin{cases} y' = \frac{dy}{dt} = f(y, t) & \text{first-order} \\ y(0) = y_0 & \text{ODE} \end{cases}$$



All methods assume that solution is known at  $0 \leq t \leq t_n$ , and use it to get sol. at  $t = t_{n+1}$ .

\* Higher-order ODEs can be converted to a system of 1st-order ODEs

$$y'' + \omega^2 y = f$$

$$y_1 = y$$

$$y_2 = y_1' = y'$$

$$y_1'' = y_2' = -\omega^2 y_1 + f$$

$$\Rightarrow \begin{cases} y_1' = y_2 \\ y_2' = -\omega^2 y_1 + f \end{cases}$$

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

Taylor series expansion

$$y_{n+1} = y(t_{n+1}) = y(t_n + \Delta t) = y_n + \Delta t y_n' + \frac{\Delta t^2}{2} y_n'' + \frac{\Delta t^3}{6} y_n''' + \dots$$

$y_n' = f(y_n, t_n)$

$$y_n'' = \left. \frac{d}{dt} y' \right|_n = \left. \frac{d}{dt} f(y, t) \right|_n = \left. \frac{\partial f}{\partial t} + \frac{\partial f}{\partial y} \frac{\partial y}{\partial t} \right|_n$$
$$= f_{t_n} + f_{y_n} f_n$$

$$y_n''' = \left. \frac{d}{dt} (f_t + f_y f) \right|_n = \left. \frac{\partial}{\partial t} (f_t + f_y f) \right|_n + \left. \frac{\partial}{\partial y} (f_t + f_y f) \frac{\partial y}{\partial t} \right|_n$$
$$= f_{t t_n} + f_{y t_n} f_n + f_{y_n} f_{t_n} + f_{t y_n} f_n + f_{y y_n} f_n f_n + f_{y_n} f_{y_n} f_n$$

# of terms increases very rapidly.

Hence, it is not very practical to include higher-order terms than third order.

① Euler method (Explicit Euler or Forward Euler)

$$y' = f(y, t) \rightarrow y_{n+1} = y_n + h f(y_n, t_n) + O(h^2)$$

$y_n \odot \rightarrow x$   
 $\frac{1}{t_n} \quad \frac{1}{t_{n+1}}$

$$y' = \lim_{h \rightarrow 0} \frac{y_{n+1} - y_n}{h} \quad \text{or} \quad \frac{y_{n+1} - y_n}{h} = f(y_n, t_n)$$

2nd-order accurate for one time step

globally (integration from  $t_0$  to  $t_f$ )

$$t_f - t_0 = nh \quad \text{error} \sim \sum h^2 y'' = nh^2 y'' = \frac{t_f - t_0}{h} \cdot h^2 y''$$

$\therefore$  1st-order accurate!

$\sim O(h)$

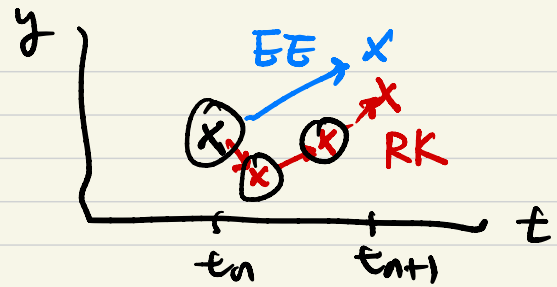
ex)  $\frac{dy}{dt} = y' = -y^2, y(0) = 1 \rightarrow$  exact sol.  $y = \frac{1}{1+t}$

$(h=0.1)$  EE  $\Rightarrow y_{n+1} = y_n + h f(y_n, t_n) = y_n - h y_n^2$   
 exact sol.

$y_0 = 1$   
 $y_1 = y_0 - h y_0^2 = 0.9$   
 $y_2 = y_1 - h y_1^2 = \dots$   
 $y_{10} = y_9 - h y_9^2 = 0.5623$       0.5

\* Runge-Kutta method  
(RK)

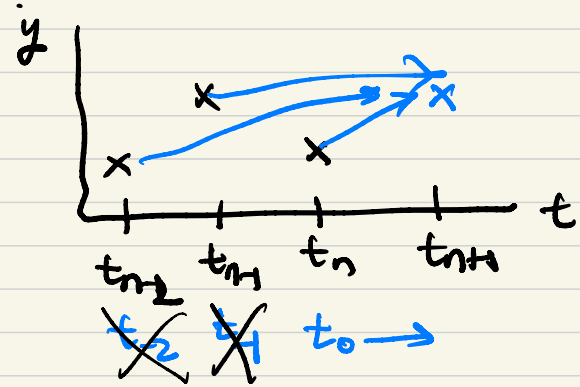
$y_{n+1}$  is obtained in terms of  $y_n$ ,  $f(y_n, t_n)$  and values of  $f$  at intermediate times ( $t_n \leq t \leq t_{n+1}$ )



\* Multi-step method  
uses information from  $t \leq t_n$

$y_{n-1}$  @  $t_{n-1}$ ,  $y_{n-2}$  @  $t_{n-2}$ , ...

not self-starting!



\* explicit vs. implicit methods



EE:  $y_{n+1} = y_n + h f(y_n, t_n)$

involves  $f(y_{n+1}, t_{n+1})$

ex) implicit Euler method

$y_{n+1} = y_n + h f(y_{n+1}, t_{n+1})$

but provides better numerical stability.

← may involve solving a nonlinear algebraic eq.

## 4.2 Numerical stability

It is possible for numerical sol. of a diff'l eq.

to blow up (grow unbounded) even though the exact sol. is well behaved.

→ We seek parameters of the numerical method such as  $h$  ( $\epsilon = O(h)$ ) so that numerical sol. is well behaved.

- Stable numerical method (A-stable numerical method)  
(absolutely stable " " )  
→ Numerical sol. does not blow up with any choice of parameters.

- Unstable numerical method

→ Numerical sol. always blows up irrespective of the choice of parameters.

• Conditionally stable numerical method

→ Numerical sol. is well behaved with some choice of parameters.

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

accuracy

• Linear stability analysis

$$y' = f(y, t) = f(y_0, t_0) + (t - t_0) \frac{\partial f}{\partial t} \Big|_{y_0, t_0} + (y - y_0) \frac{\partial f}{\partial y} \Big|_{y_0, t_0} + \text{HOT} \quad (y^2, t^2, \dots)$$

$$= \lambda y + \alpha_1 + \alpha_2 t + \text{HOT}$$

a particular sol.

(which doesn't affect stability)

usually provides a decaying sol.

Model problem

$$y' = \lambda y \quad \text{given } y(0) = y_0$$

ex)

$$y' = \alpha y^2$$

$\lambda = \lambda_R + i\lambda_I$  : complex number  
 exact sol.  $y = y_0 e^{\lambda t} = y_0 e^{\lambda_R t} e^{i\lambda_I t}$

$$\lambda_R \leq 0 \quad y = \frac{1}{a - \alpha t}$$

for stability

Midterm after finishing Ch. 4 (maybe  
some day in Nov.)

Model problem:  $y' = \lambda y$       $y' = f(y, t)$

$$\lambda = \lambda_R + i \lambda_I$$

4.3 stability analysis for Euler method

$$y' = f(y, t) \xrightarrow{EE} \frac{y_{n+1} - y_n}{h} = f(y_n, t_n)$$

$$y_{n+1} = y_n + h f(y_n, t_n)$$

Apply EE to model problem  $y' = \lambda y = f$

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

$$\rightarrow y_n = (1 + \lambda h)^n y_0 \quad (t = nh)$$

$$= (1 + \lambda_R h + i \lambda_I h)^n y_0$$

Whether sol. remains bounded depends on  $\lambda_R h$  &  $\lambda_I h$

$$y_1 = (1 + \lambda h) y_0$$

$$y_2 = (1 + \lambda h)^2 y_0$$

$\vdots$



For exact sol. to be well behaved

$$EE: y_n = (1 + \lambda_R h + i \lambda_I h)^n y_0 \\ = \sigma^n y_0$$

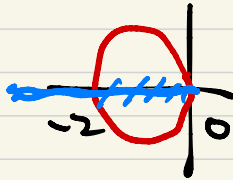
to be stable,  $|\sigma| \leq 1$

$$|\sigma|^2 = (1 + \lambda_R h)^2 + (\lambda_I h)^2 \leq 1$$

EE is stable.

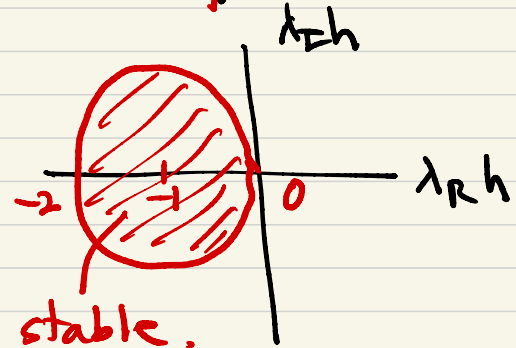
$\therefore$  EE is conditionally stable.

• when  $\lambda$  is real & negative ( $\lambda_I = 0$ )

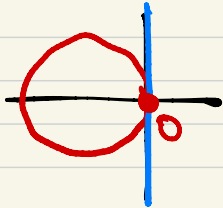


$$|\lambda h| \leq 2 \rightarrow h_{\max} = 2/|\lambda|$$

$$y' = -2y \xrightarrow{EE} h_{\max} = \frac{2}{|-2|} = 1 \text{ for stability}$$



- when  $\lambda$  is purely imaginary ( $\text{Re} = 0$ ) ( $y' = i\omega y$ )



EE is unstable for  $\lambda = i\omega$ .  $\rightarrow$  sol. blows up.

$$|S|^2 = (1 + \lambda^2 h^2)^2 + (\lambda h)^2 > 1$$

- Accuracy of EE

$y' = \lambda y$  exact sol.  $y = y_0 e^{\lambda t} = y_0 e^{\lambda n h} = y_0 \left( 1 + \lambda h + \frac{1}{2} \lambda^2 h^2 + \dots \right)^n$

EE:  $y_n = (1 + \lambda h)^n y_0$

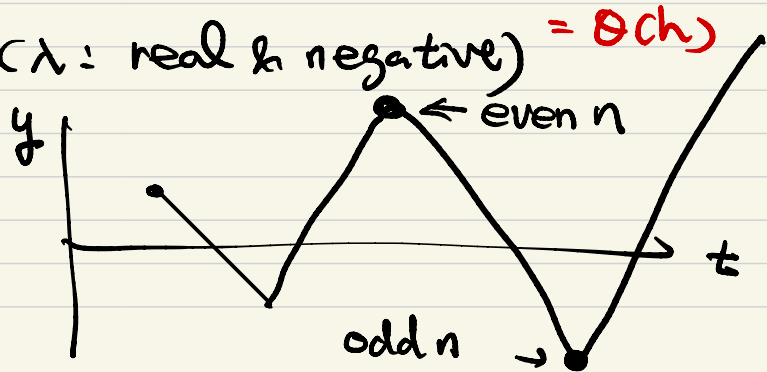
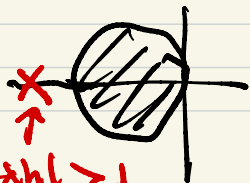
$\therefore$  EE is 1st-order accurate.

leading error

$$\frac{1}{2} \lambda^2 h^2 \cdot n = \frac{1}{2} \lambda^2 h \cdot \frac{T}{h}$$

- \* Signal for instability ( $\lambda$ : real & negative) =  $\mathcal{O}(h)$

EE:  $y_n = (1 + \lambda h)^n y_0$



#### 4.4 Implicit Euler or backward Euler method

$$y' = f(y, t) \xrightarrow{\text{IE}} \boxed{\frac{y_{n+1} - y_n}{h} = f(y_{n+1}, t_{n+1})}$$

$$y_{n+1} = y_n + h f(y_{n+1}, t_{n+1})$$

cost/timestep is higher than EE

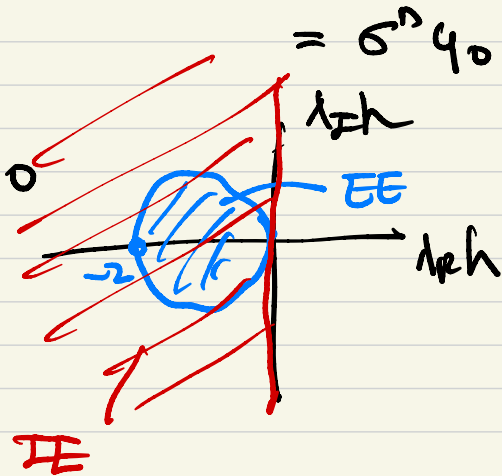
model prob.  $y' = \lambda y$

$$\text{IE: } y_{n+1} = y_n + h \lambda y_{n+1} \rightarrow y_{n+1} = \frac{1}{1 - \lambda h} y_n \rightarrow y_n = \left(\frac{1}{1 - \lambda h}\right)^n y_0$$

$$\sigma = \frac{1}{1 - \lambda h} = \frac{1}{1 - \lambda_{\text{Re}} h - i \lambda_{\text{Im}} h}$$

$$|\sigma|^2 = \frac{1}{(1 - \lambda_{\text{Re}} h)^2 + (\lambda_{\text{Im}} h)^2} \leq 1 \text{ for } \lambda_{\text{Re}} \leq 0$$

$\therefore$  IE is unconditionally stable  
A-stable



4.5

## Numerical accuracy

$$\text{IE: } \sigma = \frac{1}{1-\lambda h} = 1 + \lambda h + (\lambda h)^2 + (\lambda h)^3 + \dots$$

$$\text{exact sol.: } e^{\lambda h} = 1 + \lambda h + \frac{1}{2}(\lambda h)^2 + \frac{1}{6}(\lambda h)^3 + \dots$$

$$\text{IE: } y_n = \sigma^n y_0 = \underbrace{c(1 + \lambda h + (\lambda h)^2 + (\lambda h)^3 + \dots)}_{\text{error}} \overset{n}{y_0}$$

IE is 2nd-order accurate for one time step  
 but is 1st-order accurate overall.  
 $\lambda^2 h^2 \cdot n = \lambda^2 h^2 \cdot \frac{T}{h} = O(h)$

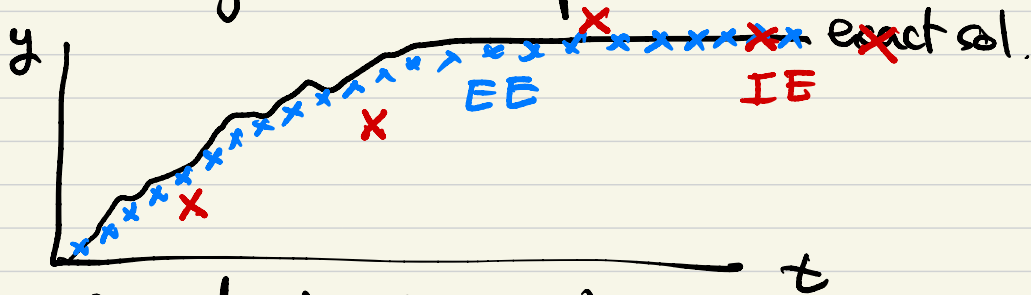
$$\text{IE: } e^{\lambda h} - \sigma = e^{\lambda h} - \frac{1}{1-\lambda h} = -\frac{1}{2}\lambda^2 h^2 + \dots$$

$$\text{EE: } e^{\lambda h} - \sigma = e^{\lambda h} - c(1 + \lambda h) = +\frac{1}{2}\lambda^2 h^2 + \dots$$

⇒ stability is nothing to do with accuracy.

From stability point of view,  
our objective is to take largest time step  $h$ ,

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



- Accuracy for  $\lambda = i\omega$  (purely imaginary)

$$y' = \lambda y = i\omega y$$

exact sol.  $y = y_0 e^{i\omega t} = y_0 (\cos \omega t + i \sin \omega t)$

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

$$\rightarrow y_n = (1 + \lambda h)^n y_0 = \sigma^n y_0 \quad \cdot \quad \sigma = 1 + i\omega h$$

exact sol. : amplitude  $|e^{i\omega t}| = 1$

EE : amplitude  $|\sigma| = \sqrt{1 + \omega^2 h^2} > 1$  unstable

$$\sigma = |\sigma| e^{i\theta} = 1 + i\omega h$$

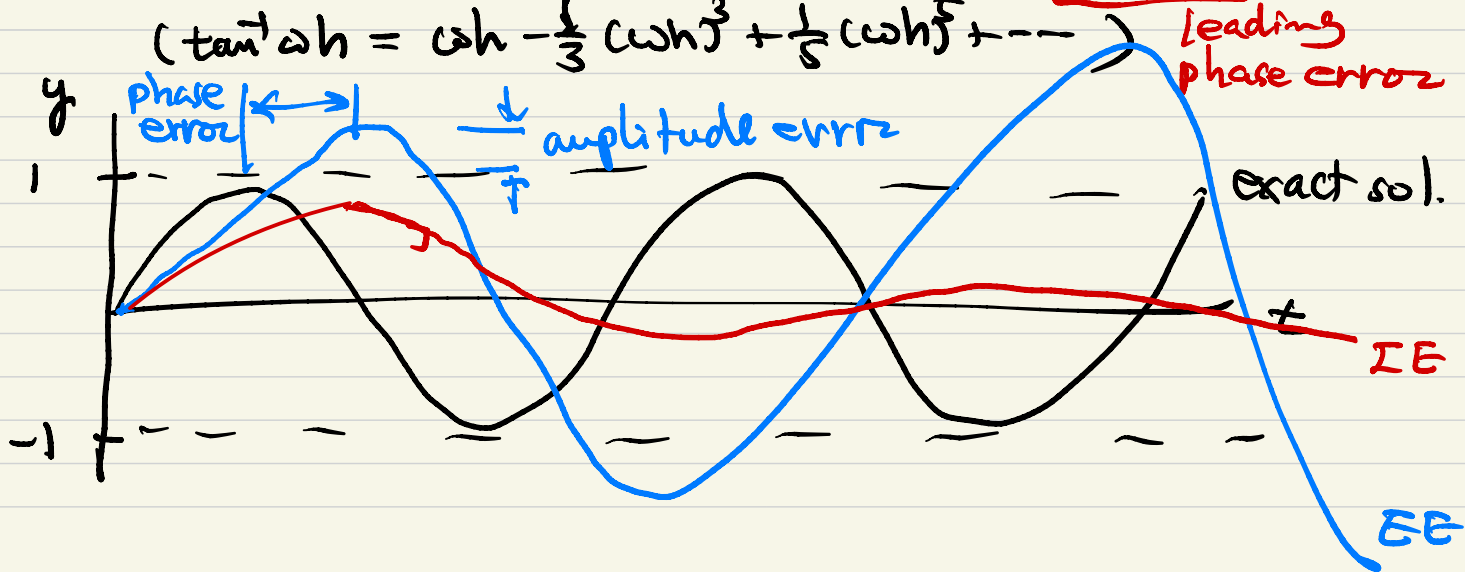
$$\text{Phase } \theta = \tan^{-1} \frac{\text{Im}(\sigma)}{\text{Re}(\sigma)} = \tan^{-1} \frac{\omega h}{1} = \tan^{-1} \omega h$$

$$\text{Exact sol. : } y = y_0 e^{i\omega t} = y_0 e^{i\omega h n} = y_0 \cdot 1 \cdot e^{i\omega h n}$$

$$\text{EE : } y_n = \sigma^n y_0 = y_0 |\sigma|^n e^{i\theta n}$$

$$\text{phase error} = \omega h - \theta = \omega h - \tan^{-1} \omega h = \frac{1}{3} (\omega h)^3 + \dots$$

$$(\tan^{-1} \omega h = \omega h - \frac{1}{3} (\omega h)^3 + \frac{1}{5} (\omega h)^5 + \dots)$$



$$\text{IE: } y_{n+1} = \frac{1}{1-i\omega h} y_n = \frac{1}{1-i\omega h} y_n \rightarrow y_n = \left(\frac{1}{1-i\omega h}\right)^n y_0$$

$$\sigma = \frac{1}{1-i\omega h} = |\sigma| e^{i\theta} = \sigma^n y_0$$

$$|\sigma|^2 = \frac{1}{1+\omega^2 h^2} < 1 \quad \text{stable} \rightarrow \text{but decaying sol.}$$

$$\sigma = \frac{1}{1-i\omega h} = \frac{1+i\omega h}{1+\omega^2 h^2}$$

$$\theta = \tan^{-1} \frac{\omega h}{1} = \tan^{-1} \omega h : \text{ same as that of EE.}$$

$$\text{phase error} = \omega h - \theta = \frac{1}{3} (\omega h)^3 + \dots$$