

Continue stability analysis of generalized alpha method

$$\left\{ \begin{array}{l} \alpha < \frac{1}{2} \text{ Conditionally stable } \Delta t \lambda < \frac{2}{1-2\alpha} \\ \alpha \geq \frac{1}{2} \text{ Unconditionally stable} \end{array} \right.$$

$\dot{u} + \lambda u = 0$  SDOF

$M \ddot{d} + K d = 0$  MDOF

$M$   $n \times n$  matrix  
 $d$  vector  
 $K$   $n \times n$  matrix

$\Rightarrow$   $\left\{ \begin{array}{l} M \text{ SDOF} \\ \ddot{d} + \lambda d = 0 \end{array} \right.$

$\rightarrow$  stability to  $\lambda$

requires

$$\left\{ \begin{array}{l} \alpha < \frac{1}{2} \\ \alpha \geq \frac{1}{2} \end{array} \right.$$

$\Delta t \leq \frac{1}{\lambda} \frac{2}{1-2\alpha}$   
 unconditionally stable

Let's assume  $\alpha < \frac{1}{2}$

$l=1$	$\Delta t \leq \frac{1}{\lambda_1} \frac{2}{1-2\alpha}$	} $\Rightarrow$
$l=2$	$\Delta t \leq \frac{1}{\lambda_2} \frac{2}{1-2\alpha}$	
$\vdots$	$\vdots$	
$l=m$	$\Delta t \leq \frac{1}{\lambda_m} \frac{2}{1-2\alpha}$	

$$z = m \quad \Delta t \leq \left( \frac{1}{\lambda m} \frac{1}{1-2\alpha} \right)$$

$$\Delta t \leq \text{Min} \left( \frac{1}{\lambda} \right) \frac{2}{1-2\alpha}$$

$$\Rightarrow \Delta t \leq \frac{1}{\text{Max}(\lambda)} \frac{2}{1-2\alpha}$$

*we need this if  $\alpha < \frac{1}{2}$*

Maximum frequency (eigenvalue) of  $M\dot{d} + Kd = 0$

$$M\dot{d} + Kd = 0$$

want to directly integrate this without do the modal analysis to find  $\text{Max}(\lambda)$

$$M \left( \frac{d_{n+1} - d_n}{\Delta t} \right) + K \left( (1-\alpha)d_{n+1} + \alpha d_n \right) = 0$$

$$\left\{ \frac{M}{\Delta t} + (1-\alpha)K \right\} d_{n+1} = \left\{ \frac{M}{\Delta t} - \alpha K \right\} d_n$$

need to solve

$$K\psi_0 = \lambda M\psi_0$$

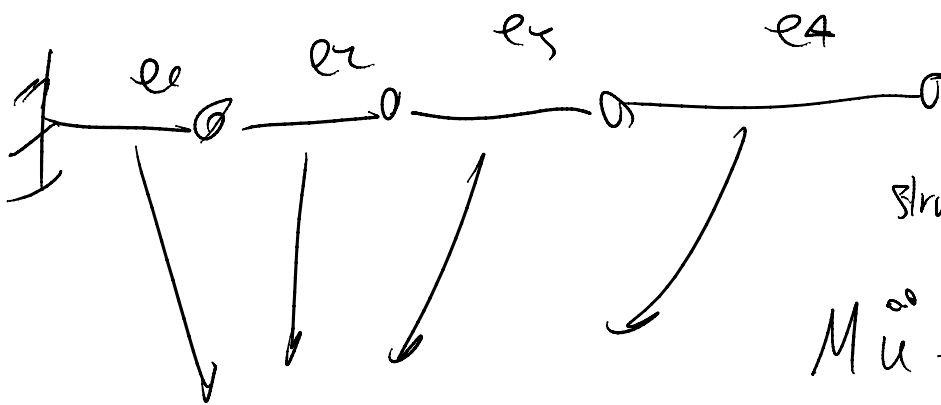
We don't actually need to do the expensive modal analysis

• Once can prove [Hughes, 2012, Bathe, 2006],

$$\lambda_0^h \geq \max_i(\lambda_i^h)$$

(311)

Maximum frequency of individual elements  
in the domain



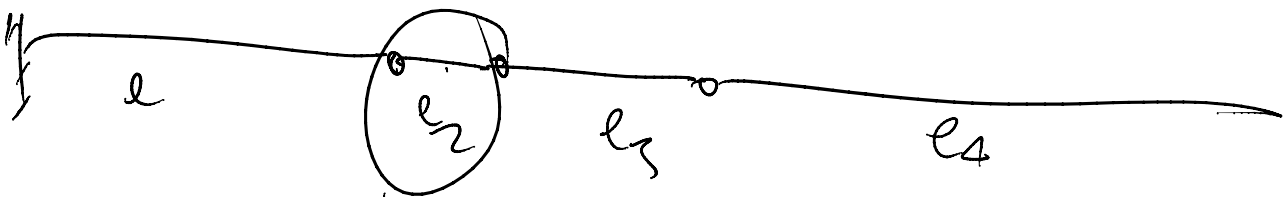
$$K_{2 \times 2}^e = \frac{AE}{L} \begin{bmatrix} 1 & -1 \\ -1 & 1 \end{bmatrix}$$

Eigenvalue problem at the element level

$$K \phi = \omega^2 M \phi$$

$$M_c^e = \frac{AL \rho A L^3}{6} \begin{bmatrix} 2 & 1 \\ 1 & 2 \end{bmatrix}$$

which element has the highest frequency



$1^{\text{st}}$   $11^{\text{th}}$   $C$

↓ has the highest frequency

$$\lambda_e^m = \sqrt{2} \frac{c}{d}$$

with consistent mass

$\lambda_e^m > \text{Max}(d)$   
 ↓  
 max freq. of individual elements  
 max freq. of MDOF expensive to calculate

$$\frac{1}{\lambda_e^m} < \frac{1}{\text{Max}(d)}$$

if  $\Delta t < \frac{1}{\lambda_e^m} \frac{2}{1-2\alpha} < \frac{1}{\text{Max}(d)} \frac{2}{1-2\alpha}$

$\Rightarrow \Delta t < \frac{1}{\text{Max}(d)} \frac{2}{1-2\alpha}$ 
→ easy to calculate

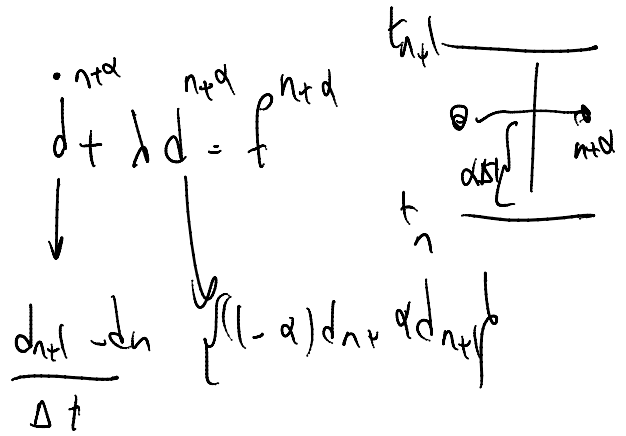
We talked about stability

Second concept is consistency

We have the numerical method update equation

$$d_{n+1} = A d_n + L_n \Rightarrow \boxed{d_{n+1} - A d_n - L_n = 0}$$

$$L_n = \Delta t \frac{(1-\alpha)f_n + \alpha f_{n+1}}{1 + \alpha \Delta t \lambda}$$



$$\frac{d_{n+1} - d_n}{\Delta t} + \lambda \left[ (1-\alpha)d_n + \alpha d_{n+1} \right] = f_{n+\alpha}$$

$$(1 + \alpha \Delta t \lambda) d_{n+1} = (1 - \lambda \Delta t (1-\alpha)) d_n + \Delta t f_n$$

$$d_{n+1} = A d_n + \underbrace{\Delta t f_n}_{L_n}$$

$$A = \frac{1 - (1-\alpha)\lambda \Delta t}{1 + \alpha \lambda \Delta t}$$

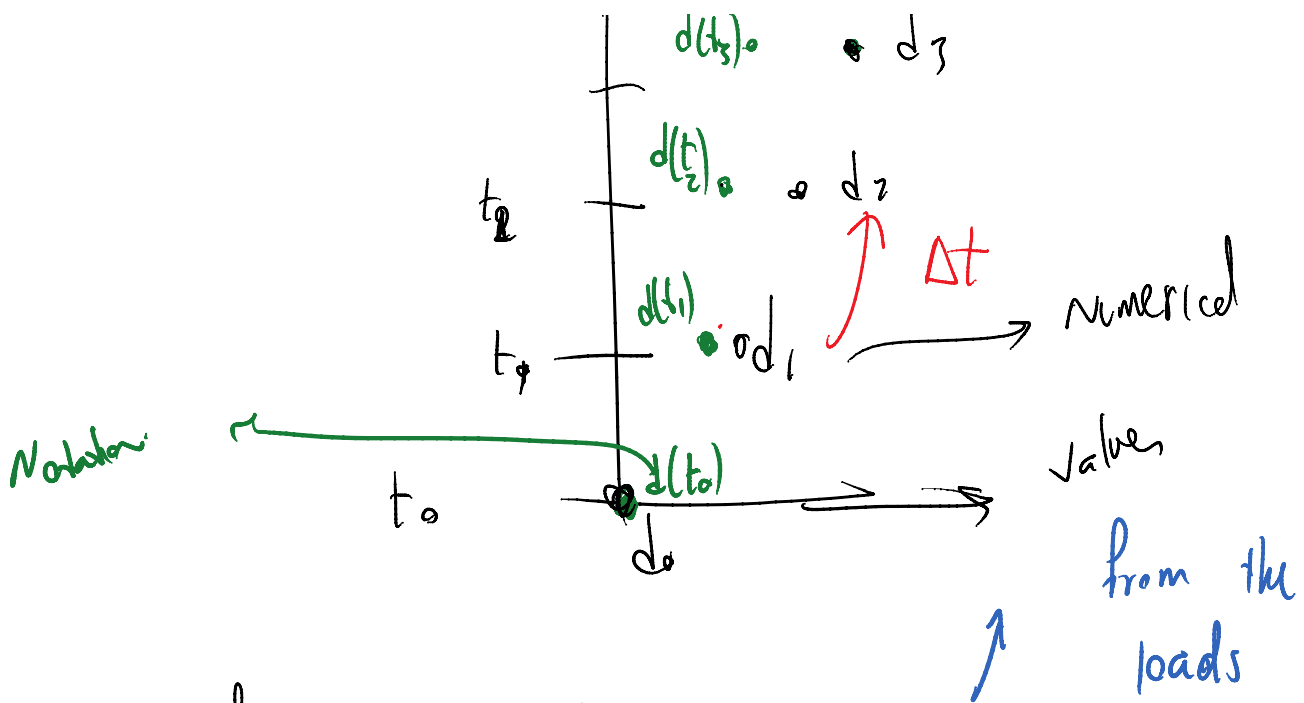
$$\boxed{d_{n+1} = A d_n + L_n}$$

$$A = \frac{1 - (1-\alpha)\lambda \Delta t}{1 + \alpha \lambda \Delta t}$$

↓ Numerical update

$$L_n = \frac{f_{n+\alpha}}{1 + \alpha \Delta t} = \frac{(1-\alpha)f_n + \alpha f_{n+1}}{1 + \alpha \Delta t}$$





Numerical

$$d_{n+1} = A d_n + L_n$$

What is the exact solution update?

$$\dot{d} + \lambda d = 0 \quad d(t_{n+1}) = d(t_n) e^{-\lambda [t_{n+1} - t_n]}$$

Consistency: is concerned with closeness

"consistency" of the numerical method & the underlying exact solution PDE at ONE

update (LOCAL update)

What is consistency for ODE's in this case  
 Generalized alpha method.

+  $\begin{cases} d_{n+1} = A d_n + L_n \end{cases}$  ↑ from lead  
 Numerical

-  $\begin{cases} d(t_{n+1}) = A d(t_n) + L_n + \Delta t \tau(t_n) \end{cases}$

$d_{n+1} - d(t_{n+1}) = A \{ d_n - d(t_n) \} - \Delta t \tau(t_n)$  truncation error

$e_{n+1} = A e_n - \Delta t \tau(t_n)$

→ Consistency requires  $\tau(t_n) \leq C \Delta t^k$

-2 Stability  $|A| \leq 1$

$k \geq 0$   
 order of accuracy

Do they result in convergence

$$\left. \begin{aligned} e_{n+1} &= A e_n - \Delta t \Sigma(t_n) \\ e_n &= A e_{n-1} - \Delta t \Sigma(t_{n-1}) \end{aligned} \right\} \Rightarrow$$

$$e_{n+1} = A (A e_{n-1} - \Delta t \Sigma(t_{n-1})) - \Delta t \Sigma(t_n)$$

$$= A^2 e_{n-1} - \Delta t A \Sigma(t_{n-1}) - \Delta t \Sigma(t_n)$$

$$e_{n-1} = A e_{n-2} - \Delta t \Sigma(t_{n-2})$$

$$e(t_{n+1}) = A^{n+1} e(t_0) - \Delta t \sum_{i=0}^n A^i \tau(t_{n-i})$$

$$n+1 \rightarrow n$$

$$e(t_n) = A^n \underbrace{e(t_0)}_{\parallel 0} - \Delta t \sum_{i=0}^{n-1} A^i \tau(t_{n-i})$$

because we start with IC

$$d_0 = d(t_0)$$



$$e(t_n) = -\Delta t \sum_{i=0}^{n-1} A^i z(t_{n-1-i})$$

$\downarrow$  stability       $\downarrow$  consistency

$$|e(t_n)| = \left| -\Delta t \sum_{i=0}^{n-1} A^i |z(t_{n-1-i})| \right| \quad |a+b| \leq |a|+|b|$$

$$\leq \Delta t \sum_{i=0}^{n-1} |A|^i |z(t_{n-1-i})|$$

$|A| \leq 1 \Rightarrow$   
 $|A|^i \leq 1$   
 stability

$$\leq \Delta t \left( \sum_{i=0}^{n-1} 1 \right) C (\Delta t)^k$$

$$= \Delta t (n) C (\Delta t)^k$$

$$C t_n (\Delta t)^k$$

$$|e(t_n)|$$

$$\leq C_T (\Delta t)^k$$

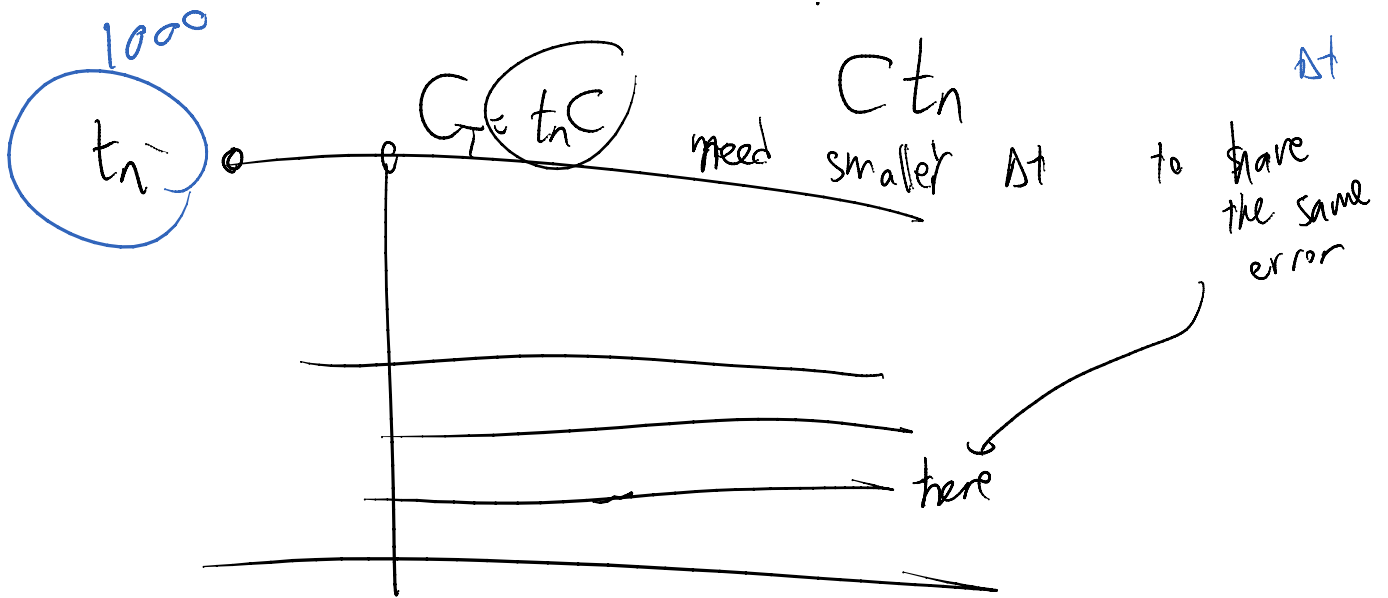
order of accuracy  
"convergence rate"

depends on time

problem would have been if  $C_T$  was dependent on  $\Delta t$

1000

$t_n$



$\lambda$  repeated 7 times  $n_{\lambda}^A = 4$

2 eigenvectors independent  $n_{\lambda}^G = 2$

