

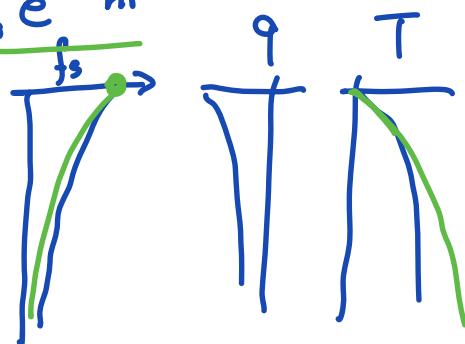
## Lecture 17 : Heat equation

Logistics: - HW 7 extension to Sat 11:45 pm

Last time: - steady heat equation

$$-\nabla \cdot [\kappa \nabla T] = \rho H_0 e^{-z/\text{hr}}$$

$\Rightarrow$  Crustal geotherm



- Approx. of  $f_s(x)$

$\Rightarrow$  best to average  $f_s$  over cell

Today: - transient heat equation

- Theta method

- Amplification matrix

- Decay of localized heat pulse

Energy balance equation

$$\bar{\rho} \bar{c}_p \frac{\partial T}{\partial t} + \nabla \cdot [q_{pf} c_{pf} T - \bar{\kappa} \nabla T] = \underline{\rho H}$$

Today:  $\phi=0 \rightarrow q=0 \quad \bar{\kappa} = \kappa_s = \kappa$

$$\bar{\rho} \bar{c}_p = \rho_s c_{pis} = \rho c_p$$

Heat equation

$$\boxed{\rho c_p \frac{\partial T}{\partial t} - \nabla \cdot [\kappa \nabla T] = \rho H} \quad u = T(\underline{x}_e)$$

new

$$\rho c_p \frac{\partial u}{\partial t} - D * [ \underline{k} \underline{d} \underline{G} u ] = f_s$$

$\Rightarrow$  just need to discretize time derivative

$$\frac{\partial u}{\partial t} = \frac{u^{n+1} - u^n}{\Delta t}$$

simple finite difference

$$t^{n+1} = t^n + \Delta t = \text{next}$$

$$t^n = n \Delta t$$

$n$  = index for time level

substitute

$$S = \underline{(u^{n+1} - u^n)} + \underline{\Delta t} \underline{\underline{L}} \underline{\underline{u}}^? = \underline{\Delta t} f_s \quad (\underline{\underline{L}} = -D \underline{k} \underline{d} \underline{G})$$

where  $\underline{\underline{S}} = \rho c_p \underline{\underline{I}}$   $\underline{\underline{I}}$   $N$  by  $N$

if  $\rho c_p = \text{const.}$

$$\underline{\underline{S}} = \begin{pmatrix} & \\ & \rho c_p & \\ & & \end{pmatrix}$$

" $\underline{\underline{S}}$ " storage  
of heat

### Theta method

decide when to evaluate  $\underline{\underline{L}} \underline{\underline{u}}$ !

$$\underline{\underline{u}}^6 = \theta \underline{\underline{u}}^n + (1-\theta) \underline{\underline{u}}^{n+1}$$

substitute:

$$\underline{\underline{S}} (\underline{\underline{u}}^{n+1} - \underline{\underline{u}}^n) + \Delta t \underline{\underline{L}} \underline{\underline{u}}^6 = [\theta \underline{\underline{u}}^n + (1-\theta) \underline{\underline{u}}^{n+1}] = \Delta t \underline{\underline{f}}_s$$

we know  $\underline{\underline{u}}^n$  we need to determine  $\underline{\underline{u}}^{n+1}$

move all known terms to r.h.s.

$$\underbrace{[\underline{\underline{S}} + \Delta t (1-\theta) \underline{\underline{L}}]}_{\text{IM}} \underline{\underline{u}}^{n+1} = \Delta t \underline{\underline{f}}_s + \underbrace{[\underline{\underline{S}} - \Delta t \theta \underline{\underline{L}}]}_{\text{EX}} \underline{\underline{u}}^n$$

Linear system for a single time step:

$$\underline{\underline{M}} \underline{\underline{u}}^{n+1} = \Delta t \underline{\underline{f}}_s + \underline{\underline{E}} \underline{\underline{X}} \underline{\underline{u}}^n$$

Implicit matrix :  $\underline{\underline{M}} = \underline{\underline{S}} + \Delta t (1-\theta) \underline{\underline{L}}$

Explicit matrix :  $\underline{\underline{E}}\underline{x} = \underline{\underline{S}} - \Delta t \theta \underline{\underline{L}}$

$\Rightarrow$  solved with `solue_Llup.m`

### Properties of the $\theta$ -method

For  $\theta=1$  : Forward Euler Method

$$\begin{aligned}\underline{\underline{M}} &= \underline{\underline{S}} + \Delta t (1-\theta) \underline{\underline{L}} = \underline{\underline{S}} \quad (\text{diagonal}) \\ \Rightarrow \underline{\underline{u}}^{n+1} &= \underline{\underline{S}}^{-1} (\Delta t f_s + \underline{\underline{E}}\underline{x} \underline{\underline{u}}^n)\end{aligned}$$

- explicit update (don't need to solve linear system)
- only matrix-vector multiply  $\rightarrow$  cheap
- conditionally stable:  $\Delta t \leq \frac{\Delta x^2}{2\kappa}$

$$\kappa = \frac{k}{\rho c_p}$$

- first-order accurate

...

For  $\theta=0$  : Backward Euler Method

$$\underline{\underline{E}}\underline{x} = \underline{\underline{S}}$$

$$\underline{\underline{M}} \underline{\underline{u}}^{n+1} = \Delta t f_s + \underline{\underline{E}}\underline{x} \underline{\underline{u}}^n$$

- implicit method

- solve linear system at every time step
- unconditionally stable
- first-order accurate

For  $\Theta = \frac{1}{2}$ : Crank - Nicolson Method

$$\underline{\underline{M}} \underline{u}^{n+1} = \Delta t \underline{f}_S + \underline{\underline{E}} \underline{X} \underline{u}^n$$

- implicit method
- solve linear system + matrix vector mult.
- unconditionally stable  
(but has oscillation limit)
- second order accurate

Why do these methods behave this way?

## Amplification Matrix

Linear system

$$\underline{\underline{M}} \underline{u}^{n+1} = \underline{\underline{E}} \underline{X} \underline{u}^n + \underline{f}_s$$

we know that without heat sources any  $T$  extremes will decay until  $T = \text{const}$  at thermal eqbri.

$$\underline{u}^{n+1} = \underbrace{\underline{\underline{M}}^{-1} \underline{\underline{E}} \underline{X}}_{\underline{\underline{A}}} \underline{u}^n$$

$\underline{\underline{A}}$  = amplification matrix

$$\underline{u}^{n+1} = \underline{\underline{A}} \underline{u}^n = \underline{\underline{A}} (\underline{\underline{A}} \underline{u}^{n-1}) = \underline{\underline{A}} \underline{\underline{A}} (\underline{\underline{A}} \underline{u}^{n-2}) = \underline{\underline{A}}^n \underline{u}^0$$

where  $\underline{u}^0$  is the initial condition

$$\boxed{\underline{u}^{n+1} = \underline{\underline{A}}^n \underline{u}^0} \quad n \in \mathbb{N}$$

To evolve in time we just keep multiplying by  $\underline{\underline{A}}$ .

Compute matrix exponential using spectral decomposition

$$\underline{\underline{A}} = \underline{\underline{Q}} \underline{\Lambda} \underline{\underline{Q}}^{-1}$$

$\underline{\underline{Q}}$  = square matrix of eigenvectors

$\underline{\Lambda}$  = diagonal matrix of eigenvalues

$$\underline{A} \underline{A} = \underline{\Lambda}^2 = (\underline{Q} \underline{\Lambda} \underline{Q}^{-1})(\underline{Q} \underline{\Lambda} \underline{Q}^{-1})$$

$$= \underline{Q} \underline{\Lambda} \underbrace{\underline{Q}^{-1} \underline{Q}}_{\underline{I}} \underline{\Lambda} \underline{Q}^{-1} = \underline{Q} \underline{\Lambda} \underline{\Lambda} \underline{Q}^{-1} = \underline{Q} \underline{\Lambda}^2 \underline{Q}^{-1}$$

$$\underline{A}^n = \underline{Q} \underline{\Lambda}^n \underline{Q}^{-1}$$

$$\underline{A}\underline{B} \neq \underline{B}\underline{A}$$

What happens when we multiply a vector by a matrix

$$\underline{A}^n \underline{u}$$

Condition for stable time integration is  
that all  $|\lambda_n| \leq 1$