Finite Difference Method

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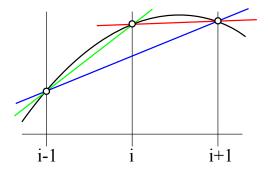


Introduction to Finite Difference Methods

- analytical solution diffucult/impossible for real problems
- finite number of compute elements
- approximation: accuracy and stability
- **4** only $u(x_i)$ known \rightarrow derivatives unknown

Introduction to Finite Difference Methods

- one-sided approximation
- centered approximation



Examples

one-sided:

$$D_{+}u(x_{i}) = \frac{u(x_{i} + h) - u(x_{i})}{h}$$
(1)

$$D_{-}u(x_{i}) = \frac{u(x_{i}) - u(x_{i} - h)}{h}$$
 (2)

centered approximation:

$$D_0 u(x_i) = \frac{u(x_i + h) - u(x_i - h)}{2h} = \frac{1}{2} (D_+ u(x_i) - D_- u(x_i))$$
(3)

third order approximation:

$$D_3u(x_i) = \frac{1}{6h}(2u(x_i+h)+3u(x_i)-6u(x_i-h)+u(x_i-2h))$$
(4)

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Truncation Error

 \bullet expand $u(x_i)$ in a Taylor series:

$$u(x_i + h) = u(x_i) + hu'(x_i) + \frac{1}{2}h^2u''(x_i) + \frac{1}{6}h^3u'''(x_i) + O(h^4)$$
 (5)

$$u(x_i - h) = u(x_i) - hu'(x_i) + \frac{1}{2}h^2u''(x_i) - \frac{1}{6}h^3u'''(x_i) + O(h^4)$$
 (6)

example: one-sided approx:

$$D_{+}u(x_{i}) = \frac{u(x_{i} + h) - u(x_{i})}{h} = u'(x_{i}) + \frac{1}{2}hu''(x_{i}) + \frac{1}{6}h^{2}u'''(x_{i}) + O(h^{3})$$
(7)

$$E(h) = D_{+}u(x_{i}) - u'(x_{i}) = \frac{1}{2}hu''(x_{i}) + \dots$$
 (8)

Truncation Error

example: centered approx:

$$u(x_i+h)-u(x_i-h)=2hu'(x_i)+\frac{1}{3}h^3u'''(x_i)+O(h^5)$$
 (9)

$$\frac{u(x_i+h)-u(x_i-h)}{2h}-u'(x_i)=\frac{1}{6}h^2u'''(x_i)+O(h^4) \quad (10)$$

Deriving Finite Difference Approximations

• example: FD approx for $u'(x_i)$ based on $u(x_i)$, $u(x_i - h)$, $u(x_i - 2h)$

$$D_2u(x_i) = au(x_i) + bu(x_i - h) + cu(x_i - 2h)$$
 (11)

Taylor:

$$D_2u(x_i) = (a+b+c)u(x_i) - (b+2c)hu'(x_i) + \frac{1}{2}(b+4c)h^2u''(x_i) \dots$$

$$\dots - \frac{1}{6}(b+8c)h^3u'''(x_i) + \dots$$
(12)

solve system:

$$a+b+c=0$$
 $b+2c=-1/n$ $b+4c=0$ (13)

resulting stencil:

$$D_2 u(x_i) = \frac{3u(x_i) - 4u(x_i - h) + u(x_i - 2h)}{2h}$$
 (14)

Deriving Finite Difference Approximations

second order derivative (example):

$$D^{2}u(x_{i}) = D_{+}D_{-}u(x_{i}) = \frac{1}{h^{2}}\left[u(x_{i} - h) - 2u(x_{i}) + u(x_{i} + h)\right]$$
 (15)

$$\cdots = u''(x) + \frac{1}{12}h^2u'''(x_i) + O(h^4)$$
 (16)

1D Elliptic Equation

- class of differential operators
- stationary problems
- minimal energy
- example: diffusion equation (isotropic)

$$u_t(x,t) = Du_{xx}(x,t) + \Psi(x,t) \tag{17}$$

- IC: $u(x,0) = u_0(x)$, BC: $u(a,t) = \alpha(t)$, $u(b,t) = \beta(t)$
- steady state with $f(x) = -\Psi(x)/D$

$$u''(x) = f(x)$$
 $0 < x < 1$ $u(0) = \alpha \ u(1) = \beta$ (18)

1D Elliptic Equation

- discretization: $\{x_0, x_1, \dots, x_{m-1}, x_m, x_{m+1}\}, h = \frac{1}{m+1}$
- unknowns: $\{x1, x2, \dots, x_m\}$
- apply second order centerd difference:

$$\frac{1}{h^2}\left[u_{i-1}-2u_i+u_{i+1}\right]=f(x_i) \qquad i=1\dots m \tag{19}$$

• write as AU = F:

$$\begin{bmatrix} -2 & 1 & & & & \\ 1 & -2 & 1 & & & \\ \dots & \dots & \dots & \dots & \dots \\ & & 1 & -2 & 1 \\ & & & 1 & -2 \end{bmatrix} \begin{bmatrix} u_1 \\ \dots \\ u_m \end{bmatrix} = \begin{bmatrix} f(x_1) - \alpha/h^2 \\ f(x_2) \\ \dots \\ f(x_{m-1}) \\ f(x_m) - \beta/h^2 \end{bmatrix}$$
(20)

Neumann Boundary Condition

- e.g. at left boundary: $u'(0) = \sigma$
- one-sided approx: $\frac{u_1-u_0}{h}=\sigma$
- centered approx:

$$\frac{1}{h^2}\left[u_{-1}-2u_0+u_1\right]=f(x_0) \tag{21}$$

$$\frac{1}{2h}[u_1 - u_{-1}] = \sigma \tag{22}$$

$$\frac{1}{h}[-u_0 + u_1] = \sigma + \frac{h}{2}f(x_0) \qquad O(h)$$
 (23)

• we want to use u_1 , u_2 , u_3 (see deriving FD approx):

$$\frac{1}{h} \left[\frac{3}{2} u_0 - 2u_1 + \frac{1}{2} u_2 \right] = \sigma \qquad O(h^2)$$
 (24)

Neumann Boundary Condition

$$\frac{1}{h^{2}} \begin{bmatrix} \frac{3}{2}h & -2h & \frac{1}{2}h & & \\ 1 & -2 & 1 & & \\ \dots & \dots & \dots & \dots & \dots \\ & & 1 & -2 & 1 \\ & & & 1 & -2 \end{bmatrix} \begin{bmatrix} u_{0} \\ u_{1} \\ \dots \\ u_{m} \end{bmatrix} = \begin{bmatrix} \sigma \\ f(x_{1}) \\ \dots \\ f(x_{m}) - \beta/h^{2} \end{bmatrix}$$
(25)

2D Elliptic Equations

- $a_1 u_{xx} + a_2 u_{xy} + a_3 u_{yy} + a_4 u_x + a_5 u_y + a_6 u = f$
- $a_2^2 4a_1a_3 < 0$
- example: Poisson equation (diffusion)

$$u_{xx} + u_{yy} = f (26)$$

- IC: $u(x, y, 0) = u_0(x, y, 0)$, BC: $u(x, y, t) \{x, y\} \in \partial \Omega$
- centerd differences:

$$\frac{1}{h_x^2}\left[u_{i-1,j}-2u_{ij}+u_{i+1,j}\right]+\frac{1}{h_y^2}\left[u_{i,j-1}-2u_{ij}+u_{i,j+1}\right]=f_{ij} \qquad (27)$$

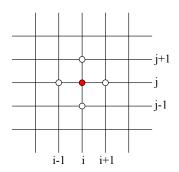
$$\frac{1}{h^2}\left[u_{i-1,j}+u_{i+1,j}+u_{i,j-1}+u_{i,j+1}-4u_{ij}\right]=f_{ij} \tag{28}$$

• lexicographic ordering: $u = [u_{11} \dots u_{m1}, u_{12} \dots u_{m2}, \dots, u_{1m} \dots u_{mm}]$

2D Elliptic Equations

centered differences:

$$\nabla_h^2 u(x) = \frac{1}{h^2} \left[u_{i-1,j} + u_{i+1,j} + u_{i,j-1} + j_{i,j+1} - 4u_{ij} \right]$$
 (29)



2D Elliptic Equations

the system now looks like:

$$A = \begin{bmatrix} T & I & & & \\ I & T & I & & & \\ & I & T & I & & \\ & & & \ddots & \ddots & \\ & & & & \ddots & \ddots \end{bmatrix} \in \mathbb{R}^{m^2 \times m^2}$$
 (30)

with

$$T = \begin{bmatrix} -4 & 1 & & & \\ 1 & -4 & 1 & & & \\ & 1 & -4 & 1 & & \\ & & & \dots & \dots \end{bmatrix} \in \mathbb{R}^{m \times m}$$
 (31)

Initial Value Problem for ODE

example:

$$u'(t) = f(u(t), t)$$
 IC: $u(t_0) = \eta$ (32)

very simple: forward Euler

$$\frac{u^{n+1} - u^n}{\Delta t} = f(u^n) \qquad O(\Delta t)$$
 (33)

less simple: backward Euler

$$\frac{u^{n+1} - u^n}{\Delta t} = f(u^{n+1}) \qquad O(\Delta t)$$
 (34)

less simple: trapezoidal method

$$\frac{u^{n+1} - u^n}{\Delta t} = \frac{1}{2} \left[f(u^n) + f(u^{n+1}) \right] \qquad O(\Delta t^2)$$
 (35)

Initial Value Problem for ODE

multistep methods: e.g.

$$\frac{u(t + \Delta t) - u(t - \Delta t)}{2\Delta t} = u'(t) + \frac{1}{6}\Delta t^2 u'''(t) + O(\Delta t^3)$$
 (36)

example: leapfrog/midpoint method

$$\frac{u(t+\Delta t)-u(t-\Delta t)}{2\Delta t}=f(u^n) \qquad O(\Delta t^2) \qquad \text{explicit 2-step}$$
(37)

- multistep methods for higher accuracy (but high memory consumption)
- accuracy can be achieved by one-step methods (e.g. Runge-Kutta)

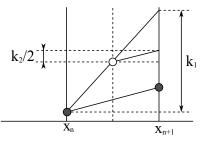
Runge-Kutta Method

- Forward Euler: $u^{n+1} = u^n + \Delta t \cdot f(x_n, u_n) O(h)$
- second order RK:
 - trial step to the midpoint of the interval
 - use info there

$$k_1 = \Delta t \cdot f(x_n, u_n) \tag{38}$$

$$k_2 = \Delta t \cdot f(x_n + \frac{1}{2}\Delta t, u_n + \frac{1}{2}k_1)$$
 (39)

$$u^{n+1} = u^n + k_2 + O(\Delta t^3)$$
 (40)



Diffusion Equation (parabolic)

$$u_t = u_{xx} (41)$$

- IC: $u(x,0) = \eta(x)$, BC: $u(0,t) = g_0(t)$, $u(1,t) = g_1(t)$
- discretization: $x_i = i \cdot \Delta x$, $t_n = n \cdot \Delta t$
- very simple: centered difference in space, forware Euler in time.

$$\frac{u_i^{n+1} - u_i^n}{\Delta t} = \frac{1}{h^2} \left[u_{i-1}^n - 2u_i^n + u_{i+1}^n \right] \qquad \frac{\Delta t}{h^2} < \frac{1}{2}$$
 (42)

Crank-Nicholson:

$$\frac{u_i^{n+1} - u_i^n}{\Delta t} = \frac{1}{2} \left[D^2 u_i^n + D^2 u_i^{n+1} \right]$$
 (43)

$$\cdots = \frac{1}{2h^2} \left[u_{i-1}^n - 2u_i^n + u_{i+1}^{n+1} - 2u_i^{n+1} + u_{i+1}^{n+1} \right]$$
 (44)

Crank-Nicholson

$$-ru_{i-1}^{n+1} + (1+2r)u_i^{n+1} - ru_{i+1}^{n+1} = ru_{i-1}^n + (1-2r)u_i^n + ru_{i+1}^n \qquad r = \frac{\Delta t}{2h^2}$$
 (45)

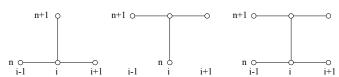
$$\begin{bmatrix} (1+2r) & -r \\ -r & (1+2r) & r \\ & -r & (1+2r) & -r \end{bmatrix} \begin{bmatrix} u_1^{n+1} \\ u_2^{n+1} \\ u_3^{n+1} \\ & \ddots \end{bmatrix} = \begin{bmatrix} r(g_0^n + g_0^{n+1}) + (1+2r)u_1^n + ru_2^n \\ ru_1^n + (1+2r)u_2^n + ru_3^n \\ & \ddots \end{bmatrix}$$

$$(46)$$

solve this system in O(m)

Stencils

stencil of forward Euler, backward Euler, Crank-Nicholson:



Order of Accuracy

lets get back to forward Euler:

$$u_i^{n+1} = u_i^n + \frac{\Delta t}{h^2} \left[u_{i-1}^n - 2u_i^n + u_{i+1}^n \right]$$
 (47)

local truncation error:

$$\tau(x,t) = \frac{u(x,t+\Delta t) - u(x,t)}{\Delta t} - \frac{1}{h^2} \left[u(x-h,t) - 2u(x,t) + u(x+h,t) \right]$$
(48)

Taylor expansion in (48):

$$\tau(x,t) = \left(u_t + \frac{1}{2}\Delta t u_{tt} + \frac{1}{6}\Delta t^2 u_{ttt} + \dots\right) - \left(u_{xx} + \frac{1}{12}h^2 u_{xxxx} + \dots\right)$$
 (49)

• recall: $u_t = u_{xx}$, $u_{tt} = u_{txx} = u_{xxxx}$

$$\tau(x,t) = \left(\frac{1}{2}\Delta t - \frac{1}{12}h^2\right)u_{xxxx} + O(\Delta t^2 + h^4)$$
 (50)

- conclusion: $O(\Delta t)$ in time, $O(\Delta h^2)$ in space
- Crank-Nicholson: $O(\Delta t^2)$ in time, $O(\Delta h^2)$ in space

Multidimensional Problem

example: 2D diffusion

$$u_t = \Delta u = u_{xx} + u_{yy}$$
 $u(x, y, 0) = \eta(x, y) \ u(x, y, t) = u_{\partial}(x, y, ts)$ (51)

discretized Laplacian:

$$\nabla_h^2 u_{ij} = \frac{1}{h^2} \left[u_{i-1,j} + u_{i+1,j} + u_{i,j-1} \mathbf{1} u_{i,j+1} - 4 u_{ij} \right]$$
 (52)

discretize in time: e.g. CN

$$\frac{u_{ij}^{n+1} - u_{ij}^{n}}{\Delta t} = \frac{1}{2} \left[\nabla_{h}^{2} u_{ij}^{n} + \nabla_{h}^{2} u_{ij}^{n+1} \right]$$
 (53)

• implicit: solve linear system in every time step

Advection Equation (hyperbolic)

- waves
- advective transport
- example:

$$u_t + au_x = 0$$
 IC: $u(x, 0) = \eta(x)$ (54)

- analytical solution: $u(x, t) = \eta(x at)$
- centered difference in space, forward Euler in time (not very stable):

$$\frac{u_j^{n+1} - u_j^n}{\Delta t} = -\frac{a}{2h} \left(u_{j+1}^n - u_{j-1}^n \right)$$
 (55)

• replace u_i^n by $\frac{1}{2}(u_{i-1}^n + u_{i+1}^n)$: Lax-Friedrichs method

$$u_{j}^{n+1} = \frac{1}{2} \left(u_{j-1}^{n} + u_{j+1}^{n} \right) - \frac{a\Delta t}{2h} \left(u_{j+1}^{n} - u_{j-1}^{n} \right) \qquad \left| \frac{a\Delta t}{h} \right| \le 1 \quad (56)$$

Literature

- Press, Teukolsky Vetterling, Flannery, Numerical Recipes, Cambridge University Press, 2007
- LeVeque, Finite Differece Methods for Ordinary and Partial Differential Equations
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- Gerald W. Recktenwald, Finite-Difference Approximations to the Heat Equation, 2011

Thanks!!

Thanks for your attention!

Slides for this talk will be available at:

http://www.bsse.ethz.ch/cobi/education

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