

Jan 16, 2004.

2.1 Constructing Difference Operators.

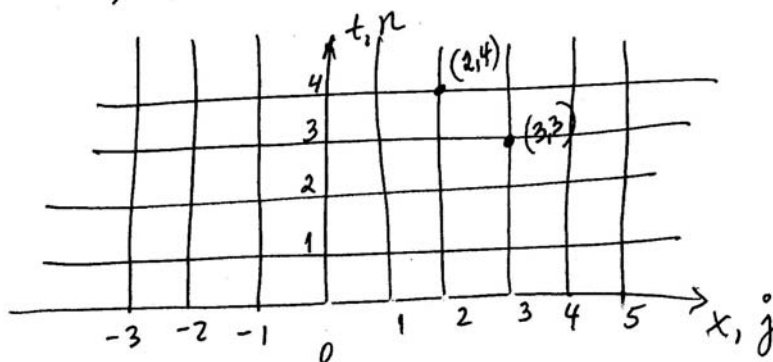
We will start constructing finite difference approxs.
or difference schemes
to 1-D problems as

$$1) \begin{cases} u_t + a u_x = 0, & -\infty < x < \infty, t > 0 \\ u(x, 0) = \phi(x) \end{cases} \quad \text{Simple wave equation}$$

$$2) \begin{cases} u_t = \sigma u_{xx}, & 0 < x < 1, t > 0 \\ u(x, 0) = \phi(x), & 0 \leq x \leq 1 \\ u(0, t) = u(1, t) = 0 \end{cases} \quad \text{Heat conduction problem.}$$

Obviously, we need finite difference approximations
for the derivative terms involved in these equations.

For example the domain for problem ① is the
upper half plane.



We make a partition as indicated:

Uniform partition of cells of size $\Delta x \times \Delta t$

These cells are obtained by making a ^{uniform} partition of the x -axis

$$x_j = j \Delta x, \quad j = \dots -3, -2, -1, 0, 1, 2, 3, \dots$$

and the t -axis: $t_n = n \Delta t, \quad n = 0, 1, 2, \dots$

The intersection of the vertical line: $x = x_j$

and the horizontal line: $t = t_n$

is the grid point: (j, n) .

And the value of the solution $u(x, t)$ at the grid

point: (x_j, t_n) or (j, n) is denoted as

$$u(x_j, t_n) = u(j \Delta x, n \Delta t) = u_j^n$$

The finite difference approx. of u_j^n is U_j^n .

Construction of Finite Difference Approximation

Keystone: Taylor's Thm.

If $f'(x), \dots, f^{(k+1)}(x)$ are defined on $[a, x]$ then

$$f(x) = f(a) + f'(a)(x-a) + \frac{f''(a)}{2}(x-a)^2 + \dots + \frac{f^{(k)}(a)}{k!}(x-a)^k + \frac{f^{(k+1)}(\eta)}{(k+1)!}(x-a)^{k+1},$$

$$\eta \in (a, x).$$

In particular, if $f(x) = u(x, t_n)$, for t_n fixed and $f(a) = u(x_j, t_n)$ and $f(x) = u(x_{j+1}, t_n)$, then $\Delta x = x_{j+1} - x_j$ and $u(x_{j+1}, t_n) = u_{j+1}^n = u_j^n + \left(\frac{\partial u}{\partial x}\right)_j^n \Delta x + \left(\frac{\partial^2 u}{\partial x^2}\right)_j^n \frac{\Delta x^2}{2} + \dots + \frac{1}{k!} \left(\frac{\partial^k u}{\partial x^k}\right)_j^n \Delta x^k + \frac{1}{(k+1)!} \left(\frac{\partial^{k+1} u}{\partial x^{k+1}}(\eta, t_n)\right) \Delta x^{k+1}$

Since $\eta \in (x_j, x_{j+1}) \Rightarrow \eta = (j+\xi)\Delta x$
with $\xi \in (0, 1)$.

↓
remainder.

(4)

If we only consider two terms and the remainder

$$u_{j+1}^n = u_j^n + \left(\frac{\partial u}{\partial x}\right)_j^n \Delta x + \frac{1}{2} \left(\frac{\partial^2 u}{\partial x^2}\right)_{j+\frac{1}{2}}^n \Delta x^2$$

Now, solving for $\left(\frac{\partial u}{\partial x}\right)_j^n$

$$\left(u_x\right)_j^n = \left(\frac{\partial u}{\partial x}\right)_j^n = \frac{u_{j+1}^n - u_j^n}{\Delta x} - \underbrace{\frac{1}{2} (u_{xx})_{j+\frac{1}{2}}^n \Delta x}_{\text{remainder}}$$

Neglecting the remainder

$$\frac{u_{j+1}^n - u_j^n}{\Delta x} = \left(u_x\right)_j^n$$

Called first forward finite difference approximation. (4.1)

Also,

$$\tau_j^n = \text{local truncation error} = -\frac{1}{2} (u_{xx})_{j+\frac{1}{2}}^n \Delta x, \quad 0 < \xi < 1.$$

and we say that (4.1) forward diffe approx. is $\mathcal{O}(\Delta x)$ "of order Δx ".

Similarly, we can obtain the formula for BACKWARD
FINITE DIFFERENCE

$$(U_x)_j^n = \frac{U_j^n - U_{j-1}^n}{\Delta x} \quad (5.1)$$

With local discretization error :

$$\tau_j^n = \frac{1}{2} (U_{xx})_{j-\alpha}^n, \quad 0 < \alpha < 1$$

CENTERED FINITE DIFFERENCE

$$U_{j+1}^n = U_j^n + (U_x)_j^n \Delta x + \frac{1}{2} (U_{xx})_j^n \Delta x^2 + \frac{1}{3!} (U_{xxx})_{j+\xi}^n \Delta x^3 \quad (5.2)$$

$$- [U_{j-1}^n = U_j^n - (U_x)_j^n \Delta x + \frac{1}{2} (U_{xx})_j^n \Delta x^2 - \frac{1}{3!} (U_{xxx})_{j-\alpha}^n \Delta x^3] \quad (5.3)$$

$$U_{j+1}^n - U_{j-1}^n = 2 (U_x)_j^n \Delta x + \frac{1}{3!} [(U_{xxx})_{j+\xi}^n + (U_{xxx})_{j-\alpha}^n] \Delta x^3$$

$$0 < \xi < 1, \quad 0 < \alpha < 1.$$

Thus, solving for $(U_x)_j^n$

$$(U_x)_j^n = \frac{U_{j+1}^n - U_{j-1}^n}{2 \Delta x} + \frac{1}{6} \left[\frac{(U_{xxx})_{j+\xi}^n + (U_{xxx})_{j-\alpha}^n}{2} \right] \Delta x^2$$

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The local truncation error

$$\tau_j^n = -\frac{1}{6} \left[\frac{(u_{xxx})_{j+\frac{1}{2}}^n + (u_{xxx})_{j-\frac{1}{2}}^n}{2} \right] \Delta x^2,$$

Using the intermediate value thm for conts. function can be written as

$$\tau_j^n = -\frac{1}{6} (u_{xxx})_{j+\beta}^n \Delta x^2, \quad (6.1) \quad -1 < \beta < 1. \quad \text{Why?}$$

Therefore, the centered finite diff approximation of the first derivative $(u_x)_j^n$ is given by

$$(u_x)_j^n = \frac{u_{j+1}^n - u_{j-1}^n}{2 \Delta x}$$

With local truncation error (6.1). We can see that the advantage of this formula is that it is of order Δx^2 , or $O(\Delta x^2)$.

Exactly the same procedure can be applied for finite diff approximations for the time derivative

$$(u_t)_j^n. \quad \text{In fact,} \quad (u_t)_j^n \approx (u_t)_j^n = \frac{u_j^{n+1} - u_j^{n-1}}{2 \Delta t}, \quad \text{Centered finite difference.}$$

(7)

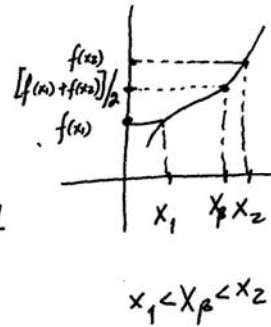
Second order centered ^{finite} difference approximation of $(u_{xx})_j^n$ can be obtained by retaining one more term in (5.2) and (5.3) and adding the resulting expressions
Do it!

$$(u_{xx})_j^n = \frac{u_{j+1}^n - 2u_j^n + u_{j-1}^n}{(\Delta x)^2}$$

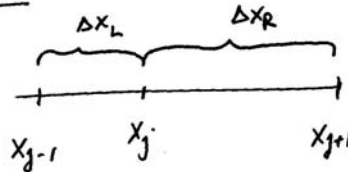
with truncation error

$$\tau_j^n = -\frac{1}{12} \left[\frac{(u_{xxxx})_{j+\beta}^n + (u_{xxxx})_{j-\alpha}^n}{2} \right] \Delta x^2$$

$$\stackrel{\text{Int. Value Thm}}{=} -\frac{1}{12} (u_{xxxx})_{j+\beta}^n, \quad -1 < \beta < 1$$



Non-uniform Grids:



Centered differences:

$$(u_x)_j^n = \frac{1}{2} \left[\frac{u_{j+1}^n - u_j^n}{\Delta x_R} + \frac{u_j^n - u_{j-1}^n}{\Delta x_L} \right] - \frac{1}{4} (\Delta x_R - \Delta x_L) (u_{xx})_j^n + O(\max(\Delta x_L^2, \Delta x_R^2))$$

Clearly, if $\Delta x_R \neq \Delta x_L$ formula is only $O(\Delta x)$
where $\Delta x = \max(\Delta x_L, \Delta x_R)$.

Example: Obtain the one-side three points formula: 7'

$$(u_x)_j = \frac{-u_{j+2} + 4u_{j+1} - 3u_j}{2\Delta x_j} + \mathcal{O}(\Delta x^2).$$

Derivation: Find $a, b,$ and c that satisfies the equation.

$$(u_x)_j = a u_{j+2} + b u_{j+1} + c u_j + \mathcal{O}(\Delta x^2).$$

$$a u_{j+2} = a \left[u_j + 2\Delta x (u_x)_j + \frac{4\Delta x^2}{2} (u_{xx})_j + \frac{8\Delta x^3}{3!} (u_{xxx})_{j+\xi} \right]$$

$$b u_{j+1} = b \left[u_j + \Delta x (u_x)_j + \frac{\Delta x^2}{2} (u_{xx})_j + \frac{(\Delta x)^3}{3!} (u_{xxx})_{j+\eta} \right]$$

$$c u_j = c u_j$$

$$\Rightarrow a u_{j+2} + b u_{j+1} + c u_j = (a+b+c)u_j + (2a+b)(u_x)_j \Delta x + \left(2a + \frac{b}{2}\right) (u_{xx})_j + \mathcal{O}(\Delta x^3).$$

$$\text{If } \begin{cases} (1) a+b+c=0 \\ (2) 2a+b=1 \\ (3) 2a+\frac{b}{2}=0 \end{cases} \Rightarrow (u_x)_j \Delta x = \frac{1}{\Delta x} [a u_{j+2} + b u_{j+1} + c u_j] + \mathcal{O}(\Delta x^2)$$

$$(2) - (3) \Rightarrow \frac{b}{2} = 1 \Rightarrow \boxed{b=2}$$

$$\Rightarrow \boxed{a = -\frac{1}{2}} \Rightarrow c = -a - b = \frac{3}{2} \Rightarrow \boxed{c = -\frac{3}{2}}$$

$$\Rightarrow (u_x)_j = \frac{-u_{j+2} + 4u_{j+1} - 3u_j}{2\Delta x_j} + \mathcal{O}(\Delta x^2) \quad \checkmark$$

$$\therefore (u_x)_j = \frac{-u_{j+2} + 4u_{j+1} - 3u_j}{2\Delta x} + O(\Delta x^2).$$

Homework: Show that

$$\textcircled{1} (u_x)_j = \frac{1}{12} [u_{j-2} - 8u_{j-1} + 8u_{j+1} - u_{j+2}] + O(\Delta x^4).$$

5 points formula.

\textcircled{2} Obtain from Table 3.1 Tannehill.
a) (3.42), b) (3.52).

Table 3.1. Comparison of formulae to evaluate $d\bar{T}/dx$ at $x=1.0$

for $\bar{T}(x) = e^x$

Case	Algebraic formula	$\left[\frac{d\bar{T}}{dx}\right]_j$	Error	Leading term in T.E.
Exact	—	2.7183	—	—
3PT SYM	$(\bar{T}_{j+1} - \bar{T}_{j-1})/2\Delta x$	2.7228	0.4533×10^{-2}	0.4531×10^{-2}
FOR DIFF	$(\bar{T}_{j+1} - \bar{T}_j)/\Delta x$	2.8588	0.1406×10^{-0}	0.1359×10^{-0}
BACK DIFF	$(\bar{T}_j - \bar{T}_{j-1})/\Delta x$	2.5868	-0.1315×10^{-0}	-0.1359×10^{-0}
3PT ASYM	$(-1.5\bar{T}_j + 2\bar{T}_{j+1} - 0.5\bar{T}_{j+2})/\Delta x$	2.7085	-0.9773×10^{-2}	-0.9061×10^{-2}
5PT SYM	$(\bar{T}_{j-2} - 8\bar{T}_{j-1} + 8\bar{T}_{j+1} - \bar{T}_{j+2})/12\Delta x$	2.7183	-0.9072×10^{-5}	-0.9061×10^{-5}

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Table 3.2. Comparison of formulae to evaluate $d^2\bar{T}/dx^2$ at $x=1.0$

Case	Algebraic formula	$\left[\frac{d^2\bar{T}}{dx^2}\right]_j$	Error	Leading term in T.E.
Exact	—	2.7183	—	—
3PT SYM	$(\bar{T}_{j-1} - 2\bar{T}_j + \bar{T}_{j+1})/\Delta x^2$	2.7205	0.2266×10^{-2}	0.2265×10^{-2}
3PT ASYM	$(\bar{T}_j - 2\bar{T}_{j+1} + \bar{T}_{j+2})/\Delta x^2$	3.0067	0.2884×10^{-0}	0.2718×10^{-0}
5PT SYM	$(-\bar{T}_{j-2} + 16\bar{T}_{j-1} - 30\bar{T}_j + 16\bar{T}_{j+1} - \bar{T}_{j+2})/12\Delta x^2$	2.7183	-0.3023×10^{-5}	-0.3020×10^{-5}

Table 3.3. Truncation error leading term (algebraic): $d\bar{T}/dx$

Case	Algebraic formula	Truncation error leading term
3PT SYM	$(\bar{T}_{j+1} - \bar{T}_{j-1})/2\Delta x$	$\Delta x^2 \bar{T}_{xxx}/6$
FOR DIFF	$(\bar{T}_{j+1} - \bar{T}_j)/\Delta x$	$\Delta x \bar{T}_{xx}/2$
BACK DIFF	$(\bar{T}_j - \bar{T}_{j-1})/\Delta x$	$-\Delta x \bar{T}_{xx}/2$
3PT ASYM	$(-1.5\bar{T}_j + 2\bar{T}_{j+1} - 0.5\bar{T}_{j+2})/\Delta x$	$-\Delta x^2 \bar{T}_{xxx}/3$
5PT SYM	$(\bar{T}_{j-2} - 8\bar{T}_{j-1} + 8\bar{T}_{j+1} - \bar{T}_{j+2})/12\Delta x$	$-\Delta x^4 \bar{T}_{xxxxx}/30$

Table 3.4. Truncation error leading term (algebraic): $d^2\bar{T}/dx^2$

Case	Algebraic formula	Truncation error leading term
3PT SYM	$(\bar{T}_{j-1} - 2\bar{T}_j + \bar{T}_{j+1})/\Delta x^2$	$\Delta x^2 \bar{T}_{xxxx}/12$
3PT ASYM	$(\bar{T}_j - 2\bar{T}_{j+1} + \bar{T}_{j+2})/\Delta x^2$	$\Delta x \bar{T}_{xxx}$
5PT SYM	$(-\bar{T}_{j-2} + 16\bar{T}_{j-1} - 30\bar{T}_j + 16\bar{T}_{j+1} - \bar{T}_{j+2})/12\Delta x^2$	$-\Delta x^4 \bar{T}_{xxxxx}/90$

Table 3-1 Difference approximations using more than three points

Derivative	Finite-difference representation	Equation
$\frac{\partial^3 u}{\partial x^3} \Big _{i,j}$	$\frac{u_{i+2,j} - 2u_{i+1,j} + 2u_{i-1,j} - u_{i-2,j}}{2h^3} + O(h^2)$	(3-38)
$\frac{\partial^4 u}{\partial x^4} \Big _{i,j}$	$\frac{u_{i+2,j} - 4u_{i+1,j} + 6u_{i,j} - 4u_{i-1,j} + u_{i-2,j}}{h^4} + O(h^2)$	(3-39)
$\frac{\partial^2 u}{\partial x^2} \Big _{i,j}$	$\frac{-u_{i+3,j} + 4u_{i+2,j} - 5u_{i+1,j} + 2u_{i,j}}{h^2} + O(h^2)$	(3-40)
$\frac{\partial^3 u}{\partial x^3} \Big _{i,j}$	$\frac{-3u_{i+4,j} + 14u_{i+3,j} - 24u_{i+2,j} + 18u_{i+1,j} - 5u_{i,j}}{2h^3} + O(h^2)$	(3-41)
$\frac{\partial^2 u}{\partial x^2} \Big _{i,j}$	$\frac{2u_{i,j} - 5u_{i-1,j} + 4u_{i-2,j} - u_{i-3,j}}{h^2} + O(h^2)$	(3-42)
$\frac{\partial^3 u}{\partial x^3} \Big _{i,j}$	$\frac{5u_{i,j} - 18u_{i-1,j} + 24u_{i-2,j} - 14u_{i-3,j} + 3u_{i-4,j}}{2h^3} + O(h^2)$	(3-43)
$\frac{\partial u}{\partial x} \Big _{i,j}$	$\frac{-u_{i+2,j} + 8u_{i+1,j} - 8u_{i-1,j} + u_{i-2,j}}{12h} + O(h^4)$	(3-44)
$\frac{\partial^2 u}{\partial x^2} \Big _{i,j}$	$\frac{-u_{i+2,j} + 16u_{i+1,j} - 30u_{i,j} + 16u_{i-1,j} - u_{i-2,j}}{12h^2} + O(h^4)$	(3-45)

Table 3-2 Difference approximations for mixed partial derivatives

Derivative	Finite-difference representation	Equation
$\frac{\partial^2 u}{\partial x \partial y} \Big _{i,j}$	$\frac{1}{\Delta x} \left(\frac{u_{i+1,j} - u_{i+1,j-1}}{\Delta y} - \frac{u_{i,j} - u_{i,j-1}}{\Delta y} \right) + O(\Delta x, \Delta y)$	(3-46)
$\frac{\partial^2 u}{\partial x \partial y} \Big _{i,j}$	$\frac{1}{\Delta x} \left(\frac{u_{i,j+1} - u_{i,j}}{\Delta y} - \frac{u_{i-1,j+1} - u_{i-1,j}}{\Delta y} \right) + O(\Delta x, \Delta y)$	(3-47)
$\frac{\partial^2 u}{\partial x \partial y} \Big _{i,j}$	$\frac{1}{\Delta x} \left(\frac{u_{i,j} - u_{i,j-1}}{\Delta y} - \frac{u_{i-1,j} - u_{i-1,j-1}}{\Delta y} \right) + O(\Delta x, \Delta y)$	(3-48)
$\frac{\partial^2 u}{\partial x \partial y} \Big _{i,j}$	$\frac{1}{\Delta x} \left(\frac{u_{i+1,j+1} - u_{i+1,j}}{\Delta y} - \frac{u_{i,j+1} - u_{i,j}}{\Delta y} \right) + O(\Delta x, \Delta y)$	(3-49)
$\frac{\partial^2 u}{\partial x \partial y} \Big _{i,j}$	$\frac{1}{\Delta x} \left(\frac{u_{i+1,j+1} - u_{i+1,j-1}}{2 \Delta y} - \frac{u_{i,j+1} - u_{i,j-1}}{2 \Delta y} \right) + O[(\Delta x, (\Delta y)^2)]$	(3-50)
$\frac{\partial^2 u}{\partial x \partial y} \Big _{i,j}$	$\frac{1}{\Delta x} \left(\frac{u_{i,j+1} - u_{i,j-1}}{2 \Delta y} - \frac{u_{i-1,j+1} - u_{i-1,j-1}}{2 \Delta y} \right) + O[(\Delta x, (\Delta y)^2)]$	(3-51)
$\frac{\partial^2 u}{\partial x \partial y} \Big _{i,j}$	$\frac{1}{2 \Delta x} \left(\frac{u_{i+1,j+1} - u_{i+1,j-1}}{2 \Delta y} - \frac{u_{i-1,j+1} - u_{i-1,j-1}}{2 \Delta y} \right) + O[(\Delta x)^2, (\Delta y)^2]$	(3-52)
$\frac{\partial^2 u}{\partial x \partial y} \Big _{i,j}$	$\frac{1}{2 \Delta x} \left(\frac{u_{i+1,j+1} - u_{i+1,j}}{\Delta y} - \frac{u_{i-1,j+1} - u_{i-1,j}}{\Delta y} \right) + O[(\Delta x)^2, \Delta y]$	(3-53)
$\frac{\partial^2 u}{\partial x \partial y} \Big _{i,j}$	$\frac{1}{2 \Delta x} \left(\frac{u_{i+1,j} - u_{i+1,j-1}}{\Delta y} - \frac{u_{i-1,j} - u_{i-1,j-1}}{\Delta y} \right) + O[(\Delta x)^2, \Delta y]$	(3-54)

Hornbeck (1975)

	f_{i-2}	f_{i-1}	f_i	f_{i+1}	f_{i+2}	
$2hf'(x_i) =$		-1	0	1		
$h^2f''(x_i) =$		1	-2	1		+ $O(h)^2$
$2h^3f'''(x_i) =$	-1	2	0	-2	1	
$h^4f^{(4)}(x_i) =$	1	-4	6	-4	1	

(a) Representations of $O(h)^2$

	f_{i-3}	f_{i-2}	f_{i-1}	f_i	f_{i+1}	f_{i+2}	f_{i+3}	
$12hf'(x_i) =$		1	-8	0	8	-1		
$12h^2f''(x_i) =$		-1	16	-30	16	-1		+ $O(h)^4$
$8h^3f'''(x_i) =$	1	-8	13	0	-13	8	-1	
$6h^4f^{(4)}(x_i) =$	-1	12	-39	56	-39	12	-1	

(b) Representations of $O(h)^4$

Fig. 3.4 Central difference representations.

	f_i	f_{i+1}	f_{i+2}	f_{i+3}	f_{i+4}	f_{i+5}	
$2hf'(x_i) =$	-3	4	-1				
$h^2f''(x_i) =$	2	-5	4	-1			+ $O(h)^2$
$2h^3f'''(x_i) =$	-5	18	-24	14	-3		
$h^4f^{(4)}(x_i) =$	3	-14	26	-24	11	-2	

(a) Forward difference representations

	f_{i-3}	f_{i-4}	f_{i-5}	f_{i-2}	f_{i-1}	f_i	
$2hf'(x_i) =$				1	-4	3	
$h^2f''(x_i) =$			-1	4	-5	2	+ $O(h)^2$
$2h^3f'''(x_i) =$		3	-14	24	-18	5	
$h^4f^{(4)}(x_i) =$	-2	11	-24	26	-14	3	

(b) Backward difference representations

Fig. 3.3 Forward and backward difference representations of $O(h)^2$.

OPERATOR NOTATION: FINITE DIFFERENCE OPERATORS.

Assuming $u(x)$

Forward Diffce	Δ	$\Delta u_j = u_{j+1} - u_j$
Backward Diffce	∇	$\nabla u_j = u_j - u_{j-1}$
Central Diffce	δ	$\delta u_j = u_{j+1/2} - u_{j-1/2}$
Average	μ	$\mu u_j = (u_{j+1/2} + u_{j-1/2})/2$
Shift	E	$E u_j = u_{j+1}$
Derivative	D	$D u_j = (u_x)_j$

Example: Centered difference formula in terms of difference operators.

$$\frac{(\mu\delta)(u_j)}{\Delta x} = \frac{u_{j+1} - u_{j-1}}{2\Delta x}$$

In fact,

$$\begin{aligned} \frac{1}{\Delta x} \mu(\delta(\mu j)) &= \frac{1}{\Delta x} \mu(u_{j+1/2} - u_{j-1/2}) = \\ &= \frac{1}{\Delta x} [\mu(u_{j+1/2}) - \mu(u_{j-1/2})] = \\ &= \frac{1}{\Delta x} \left[\frac{u_{j+1} + u_j}{2} - \frac{u_j + u_{j-1}}{2} \right] = \frac{u_{j+1} - u_{j-1}}{2\Delta x} \quad \checkmark \end{aligned}$$

(9)

Example:

$$\frac{\delta^2(u_j)}{\Delta x^2} = \frac{u_{j+1} - 2u_j + u_{j-1}}{(\Delta x)^2} \approx (u_{xx})_j$$

In fact,

$$\begin{aligned} \delta^2(u_j) &= \delta(\delta(u_j)) = \delta(u_{j+1/2} - u_{j-1/2}) = \\ &= \delta(u_{j+1/2}) - \delta(u_{j-1/2}) = u_{j+1} - u_j - u_j + u_{j-1} = \\ &= u_{j+1} - 2u_j + u_{j-1} \quad \checkmark \end{aligned}$$

Also,

$$E(u_j) = u_{j+1} = e^{\Delta x D} u_j$$

In fact,

$$\begin{aligned} u_{j+1} &= u_j + \Delta x (u_x)_j + \frac{\Delta x^2}{2} (u_{xx})_j + \dots \\ &= u_j + \Delta x D u_j + \frac{\Delta x^2}{2} D^2 u_j + \dots \\ &= (1 + \Delta x D + \frac{\Delta x^2}{2} D^2 + \dots) u_j = e^{\Delta x D} u_j \end{aligned}$$

\Rightarrow $E = e^{\Delta x D}$ (9.1)

and

$$\begin{aligned} \Delta u_j &= u_{j+1} - u_j = E u_j - u_j = (E - 1) u_j \\ \Rightarrow \quad &\boxed{\Delta = E - 1} \Rightarrow \boxed{E = 1 + \Delta} \end{aligned} \quad (9.2)$$

From (9.1) $\ln(E) = \Delta x D.$

So $\ln(E) = \ln(1+\Delta) = \Delta - \frac{1}{2}\Delta^2 + \frac{1}{3}\Delta^3 - \dots$

$\ln(1+x) = x - \frac{1}{2}x^2 + \frac{1}{3}x^3 + \dots, \quad |x| < 1.$

Defining E^{-1} , $E^{-1}(u_j) = u_{j-1}$

$\Rightarrow \nabla = 1 - E^{-1} \Rightarrow E^{-1} = 1 - \nabla$

and $\Delta x D = \ln(E) = -\ln(E^{-1}) = -\ln(1 - \nabla) =$
 $= -\left[-\nabla - \frac{1}{2}\nabla^2 - \frac{1}{3}\nabla^3 - \dots\right] = \nabla + \frac{1}{2}\nabla^2 + \frac{1}{3}\nabla^3 + \dots$

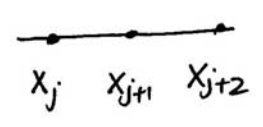
All this formal operators algebra can be proved.

Application of difference operators:

One of the application is to obtain formulas of higher order easily.

Example: One side second order finite difference approximation for $(u_x)_j$.

$(u_x)_j \approx \frac{-u_{j+2} + 4u_{j+1} - 3u_j}{2\Delta x}$ (10.1)



$T_j = O(\Delta x^2).$

(11)

Formula (10.1) can be obtained from

$$\Delta x D(u_j) \approx \ln(E)(u_j) = \ln(1 + \Delta)(u_j) =$$

$$\stackrel{\text{only 2 terms}}{\approx} \left(\Delta - \frac{1}{2} \Delta^2 \right) (u_j) = u_{j+1} - u_j - \frac{1}{2} [\Delta(u_{j+1}) - \Delta(u_j)]$$

$$= u_{j+1} - u_j - \frac{1}{2} [u_{j+2} - u_{j+1} - u_{j+1} + u_j] =$$

$$= -\frac{1}{2} u_{j+2} + 2u_{j+1} - \frac{3}{2} u_j$$

Thus,

$$(u_x)_j = D u_j = \frac{-u_{j+2} + 4u_{j+1} - 3u_j}{2 \Delta x}$$