

Science by Britanny Bahl and Amy McGovern - University of Oklahoma
Visualization by Greg Foss and Greg Abram, TACC
1500x1500x50 domain visualized with VisIt

Atms 502, CSE 566
Numerical Fluid Dynamics

TUE., FEB. 26, 2019

<https://www.tacc.ou.edu/scientific-visualization-gallery/#streamwise-vorticity-in-a-supercell-thunderstorm>

ATMS 502 - Spring 2019 2/26/19

2

ATMS 502
CSE 566

Tuesday,
26 February 2019
Class #13

- Pgm3 due Mar. 5

Plan for Today

- 1) Approximating derivatives
 - Back to Taylor series
- 2) Time differencing; Leapfrog
 - Computational *molecule*
 - Leapfrog method • stability analysis
- 3) Nesting, continued
 - Boundary conditions
 - 1-D view: interpolation, feedback
 - Grid refinement & clustering

HANDOUTS:
• SKAMAROCK & KLEMP - AMR
• SKAMAROCK - KE SPECTRA & RESOLUTION

ATMS 502 - Spring 2019 2/26/19

3

Approximating derivatives

- We have used **Taylor series** for truncation error...
- We can use the same series to **derive** approximations for derivatives

ATMS 502 - Spring 2019 2/26/19

4

Approximations to derivatives

- Consider the first derivative in space -

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

○ (Δx) ○ (Δx) ○ (Δx^2)

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

○ (Δx^4)

Following Wilhelmson

- Why *not* use higher-order approximations?
- When would you need **1-sided** approximations?

ATMS 502 - Spring 2019 2/26/19

Approximating derivatives: Example

5

- **Example: centered 1st derivative**
- Let: $\frac{df}{dx} = f_x = [af(x - \Delta x) + bf(x) + cf(x + \Delta x)]$
- Insert Taylor series expansion; get 3 equations, 3 unknowns ...
- Result: $\frac{df}{dx} = (a + b + c)f(x) + (c - a)\Delta x f'_x + (a + c)\frac{\Delta x^2}{2!} f''_{xx}$

$$a = -1/2\Delta x, b = 0, c = 1/2\Delta x \therefore f'(x) \approx \frac{f(x + \Delta x) - f(x - \Delta x)}{2\Delta x}$$

ATMS 502 - Spring 2019

2/26/19

Time differencing overview

6

SOME NOTATION AND BASIC IDEAS.

ATMS 502 - Spring 2019

C032: Operator notation for finite differences

2/26/19

Time differencing; computational diagrams

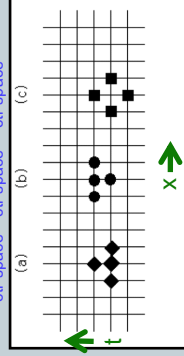
7

$$\begin{aligned} \text{a) } \frac{s_j^{n+1} - s_j^n}{\Delta t} + c \frac{s_{j+1}^n - s_{j-1}^n}{2\Delta x} &= 0 \\ \text{b) } \frac{s_j^{n+1} - s_j^n}{\Delta t} + c \frac{s_{j+1}^{n+1} - s_{j-1}^{n+1}}{2\Delta x} &= 0 \\ \text{c) } \frac{s_j^{n+1} - s_j^{n-1}}{2\Delta t} + c \frac{s_{j+1}^n - s_{j-1}^n}{2\Delta x} &= 0 \end{aligned}$$

One-sided X difference: _____
 Centered X difference: _____
 Two time levels: _____
 Three time levels: _____

Computational molecule

Fwd time, Back time, Ctr time,
 ctr space ctr space



Williamson

Explicit scheme: _____
 Implicit scheme: _____

ATMS 502 - Spring 2019

C002: Explicit methods; C003: Time levels; C004: Numerical stencil

2/26/19

The leap frog method

8

A 3-TIME-LEVEL SCHEME

Reference pages for this section:

- C003 – time levels
- C004 – numerical stencil
- C018 – complex numbers
- C020 – von Neumann analysis
- C052 – advection methods
- C054 – time differencing
- C055 – computational modes
- C056 – systems of linear equations
- C057 – eigenvalue problems
- C058 – characteristic equation
- C059 – time filtering

ATMS 502 - Spring 2019

C052: Advection methods

2/26/19

Leapfrog

⑨

- The **Leap Frog** method gets its name due to the way it makes use of data among 3 time levels.

ATMS 502 - Spring 2019 2/26/19

Leapfrog

⑩

- Leap frog method:
$$\frac{u_j^{n+1} - u_j^{n-1}}{2\Delta t} = -c \frac{u_{j+1}^n - u_{j-1}^n}{2\Delta x}$$
- Computational molecule

- Time levels
 - evaluate space derivative at time n . Store $n, n-1$ time levels.

ATMS 502 - Spring 2019 C003: Time levels • C004: Numerical stencil • C054: Time differencing 2/26/19

Leapfrog - application

⑪

- Because Leapfrog has 3 time levels ...
 - Considerations:
 - Need "help" to get started: $u_2 = f(u_1)$
 - The first time step uses a 2-time-level method
 - Applying:
 - $u_3(j) = u_1(j) - v^*(u_2(j+1) - u_2(j-1))$
 - Update: copy u_2 to u_1 ; copy u_3 to u_2
 - u_2 now contains latest results. Repeat.

ATMS 502 - Spring 2019 C003: Time levels • C054: Time differencing 2/26/19

Leapfrog - overview

⑫

- Leap frog method:
$$\frac{u_j^{n+1} - u_j^{n-1}}{2\Delta t} = -c \frac{u_{j+1}^n - u_{j-1}^n}{2\Delta x}$$
- Accuracy
 - Leapfrog is consistent and accurate of order $O[(\Delta t)^2, (\Delta x)^2]$.
 - There is considerable phase error. $u_i + cu_x = -\frac{(\Delta t)^2}{3!} u_{iii} - c \frac{(\Delta x)^2}{3!} u_{xxx} + \dots$
- Stability
 - Leapfrog is stable for $|v| \leq 1$
- Modes
 - There are two solutions from Leapfrog
 - These are the *physical* and *computational modes*
 - Results from the additional time level (higher accuracy in time)
 - A major drawback! Solution = sum of modes; comp. mode *undamped*

ATMS 502 - Spring 2019 C052: Advection methods • C055: Computational modes 2/26/19

Leapfrog - stability (13)

- Leap frog method:

$$\begin{pmatrix} u_j^{n+1} \\ v_j^{n+1} \end{pmatrix} = \begin{pmatrix} u_j^{n-1} - \mu(u_{j+1}^n - u_{j-1}^n) \\ v_j^n \end{pmatrix}$$
- Start out with usual Von Neumann method:

$$\tilde{u}^{n+1} = \tilde{u}^{n-1} - \mu \tilde{u}^n (e^{i\lambda \Delta x} - e^{-i\lambda \Delta x})$$
- Introduce new variable - "looks" 2-time-level:

Let $\tilde{v}^n = \tilde{u}^{n-1}$, so $\tilde{v}^{n+1} = \tilde{u}^n$, and we have:

$$\begin{pmatrix} \tilde{u}^{n+1} \\ \tilde{v}^{n+1} \end{pmatrix} = \begin{pmatrix} \tilde{v}^n - \mu \tilde{u}^n (2i \sin \beta) \\ \tilde{u}^n \end{pmatrix}$$

ATMS 502 - Spring 2019 C020: von Neumann stability analysis 2/26/19

Leapfrog - stability (2) (14)

- We introduced a new variable \mathbf{v} --

Let $\tilde{v}^n = \tilde{u}^{n-1}$, so $\tilde{v}^{n+1} = \tilde{u}^n$, and we have:

$$\begin{pmatrix} \tilde{u}^{n+1} \\ \tilde{v}^{n+1} \end{pmatrix} = \begin{pmatrix} \tilde{v}^n - \mu \tilde{u}^n (2i \sin \beta) \\ \tilde{u}^n \end{pmatrix}$$
- Now write in matrix form.

$$\begin{pmatrix} \tilde{u}^{n+1} \\ \tilde{v}^{n+1} \end{pmatrix} = \begin{pmatrix} \tilde{v}^n - \mu \tilde{u}^n (2i \sin \beta) \\ \tilde{u}^n \end{pmatrix} \text{ so } \begin{pmatrix} \tilde{u}^{n+1} \\ \tilde{v}^{n+1} \end{pmatrix} = \begin{pmatrix} -2i\mu \sin \beta & 1 \\ 1 & 0 \end{pmatrix} \begin{pmatrix} \tilde{u}^n \\ \tilde{v}^n \end{pmatrix}$$

ATMS 502 - Spring 2019 2/26/19

Leapfrog stability (3) (15)

- Linear algebra: $\mathbf{y} = \mathbf{A}\mathbf{x}$, $\mathbf{A}\mathbf{x} = \lambda \mathbf{x}$, $(\mathbf{A} - \lambda \mathbf{I})\mathbf{x} = \mathbf{0}$.
 - We were here:

$$\begin{pmatrix} \tilde{u}^{n+1} \\ \tilde{v}^{n+1} \end{pmatrix} = \begin{pmatrix} -2i\mu \sin \beta & 1 \\ 1 & 0 \end{pmatrix} \begin{pmatrix} \tilde{u}^n \\ \tilde{v}^n \end{pmatrix}$$
 - For a nontrivial solution, the characteristic determinant $\det(\mathbf{A} - \lambda \mathbf{I}) = 0$:

$$\begin{vmatrix} -2i\mu \sin \beta - \lambda & 1 \\ 1 & 0 - \lambda \end{vmatrix} = 0 \tag{1}$$
 - (1) is the *characteristic equation* corresponding to our matrix (A).

ATMS 502 - Spring 2019 C056: Systems of linear equations • C057: Eigenvalue problems • C058: Characteristic equation 2/26/19

Leapfrog stability (4) (16)

- Characteristic equation:

$$\begin{vmatrix} -2i\mu \sin \beta - \lambda & 1 \\ 1 & 0 - \lambda \end{vmatrix} = 0$$
- Solve:

$$\lambda^2 + 2i\mu \sin \beta \lambda - 1 = 0$$
- Two roots: things are getting interesting.

$$\lambda = -i\mu \sin \beta \pm \sqrt{1 - \mu^2 \sin^2 \beta} = -ip \pm \sqrt{1 - p^2}$$
 - If the square root is real, $|\lambda|^2 = 1$ and $|\mu| \leq 1$.
 - If the square root is imaginary, $|\lambda| > 1$.
 - Our stability condition is: $|\mu| \leq 1$

ATMS 502 - Spring 2019 C056: Systems of linear equations • C057: Eigenvalue problems • C058: Characteristic equation 2/26/19

Leapfrog - modes (17)

- We have **two modes to the solution**.
- This comes from the \pm below.

$$\lambda = -i\mu \sin \beta \pm \sqrt{1 - \mu^2 \sin^2 \beta}$$

- One is real (**physical**). One is **computational**.

Leapfrog - modes (18)

- We have **two modes to the solution**.
- This comes from the \pm below.

$$\lambda = -i\mu \sin \beta \pm \sqrt{1 - \mu^2 \sin^2 \beta}$$

- One is real (**physical**). One is **computational**.
- To find out which is which, take $\Delta t, \Delta x \rightarrow 0$.
 - ✦ Then μ goes to 0; one root goes to +1, one to -1.
 - ✦ Root of +1 is **physical**; *no growth*, all is well.
 - ✦ Root of -1: **switches sign every time step** (λ^n).
- This is a not-so-good consequence of our 3-time-level numerical scheme.

Leapfrog stability - review (19)

- We rewrote the 3-level scheme as 2-level:

$$\begin{cases} \tilde{u}^{n+1} = \tilde{v}^n - i\tilde{u}^n (2i \sin \beta) \\ \tilde{v}^{n+1} = \tilde{u}^n \end{cases} \text{ so } \begin{pmatrix} \tilde{u}^{n+1} \\ \tilde{v}^{n+1} \end{pmatrix} = \begin{pmatrix} -2i\mu \sin \beta & 1 \\ 1 & 0 \end{pmatrix} \begin{pmatrix} \tilde{u}^n \\ \tilde{v}^n \end{pmatrix}$$
- Write above as matrix, subtract I from diagonal, set determinant to zero. Characteristic equation:

$$\begin{vmatrix} -2i\mu \sin \beta - \lambda & 1 \\ 1 & 0 - \lambda \end{vmatrix} = 0$$
- Solve; 2 roots; physical and computational modes

$$\lambda = -i\mu \sin \beta \pm \sqrt{1 - \mu^2 \sin^2 \beta} = -ip \pm \sqrt{1 - p^2}$$
- As Δt and $p \rightarrow 0$: " root approaches 1, " root: -1
 - $|\lambda| = -1$ means amplitude varies as $(-1)^n$

Nesting (20)

Nested grid BCs

- Nested grid:**
- Shown below: grid-1 time step, q_1 to q_2
- Added: nested grid step, refinement factor

2

$$q_{1,grid2} = (1 - F)q_{1,grid1} + F \cdot q_{2,grid1}$$

where $F = \frac{(nstep_{grid2} - 1)}{\# steps_{grid2}}$

Time-interpolated boundary conditions

ATMS 502 - Spring 2019 2/26/19

Interpolation

- Interpolation: coarse \Rightarrow nested**

22

NESTED GRID SIZE ... EXAMPLE

- $nx=121$ (both coarse and nested grids)
- nested grid is 121 nest points wide, and
- nested grid is $(121-1)/4=30$ coarse points wide

ATMS 502 - Spring 2019 2/26/19

Interpolation

- Interpolation: coarse \Rightarrow nested**

23

NESTED GRIDS: INTERPOLATION

$$f(x) = f(a) + [f(b) - f(a)] \left(\frac{x-a}{b-a} \right)$$

ATMS 502 - Spring 2019 2/26/19

Interpolation

- Interpolation: coarse \Rightarrow nested**

24

NESTED GRIDS: INTERPOLATION

$$f(x) = f(a) + [f(b) - f(a)] \left(\frac{x-a}{b-a} \right)$$

$$\text{nest}(\text{inest}) = \text{coarse}(\text{icoarse}) + (\text{coarse}(\text{icoarse}+1) - \text{coarse}(\text{icoarse})) * \text{fraction}$$

ATMS 502 - Spring 2019 2/26/19

Interpolation

(25)

• *Interpolation: coarse ⇒ nested*

J=4 J=1

5 6 7 8 9 10 11 12 13 17 21 25 37

THIS EXAMPLE: 3:1 NESTING

NESTED GRIDS: COORDINATES

- Example above: nested grid index $j=1$
... is at coarse grid index $J=6$
- $i_{\text{coarse}} = (i_{\text{nest}} - 1) / \text{ratio} + \text{first.nest.point}$

ATMS 502 - Spring 2019 2/26/19

Interpolation

(26)

• *Interpolation: old vs. new nested grids*

Old nest new nest

THIS EXAMPLE: 3:1 NESTING

- Nested grid re-location:
 - ✓ interpolate from coarse ⇒ new nested grid
 - ✓ copy overlap region of old nest to new nest

ATMS 502 - Spring 2019 2/26/19

Interpolation

(27)

How often should we relocate the nest?

- As often as possible?
 - Minimizes copying coarse data ⇒ nest
 - This is more computationally expensive
 - ✖ Imagine computing the truncation errors over a larger 2d domain, frequently
- As rarely as possible?
 - Eventually features of interest leave the nest
 - Much of new nest would then be copied from coarse grid

ATMS 502 - Spring 2019 2/26/19

Feedback

(28)

• *Feedback: copy nested ⇒ coarse*

J=4 J=1

5 6 7 8 9 10 11 12 13 17 21 25 37

UPDATING THE COARSE GRID

- We require every 4th nest point to overlap a coarse point
- In feedback, every 4th interior nested grid point is copied back to the coarse grid.
- What alternate approach might we consider ??

ATMS 502 - Spring 2019 2/26/19

Grid refinement & Clustering

29

ADAPTIVE MESH REFINEMENT

Reference pages for this section:

- C008 – Truncation error
- C009 – Resolution
- C010 – AMR / nesting
- C051 – Nesting: grid placement, movement

ATMS 502 - Spring 2019

2/26/19

Regridding procedure

30

- **Regridding** (Skamarock dissertation, pp. 12-13)
 - 1) flag points needing refinement
 - ✦ flagged if estimated error exceeds user threshold
 - 2) cluster the flagged points – for two reasons
 - ✦ a) separates “spatially distinct phenomena [like] shocks or fronts”
 - ✦ b) subdivide to use several grids instead of one large region
 - 3) fit rectangular grids around the clustered points
 - 4) repeat steps 2,3, using different methods if necessary
 - ✦ simple method [nearest neighbor] – ok for clustering, not rectangles
 - ✦ connecting points – use *minimum spanning trees* or *nearest neighbor graphs*
 - clustering, fitting rectangles “most difficult part of regridding”

ATMS 502 - Spring 2019

2/26/19

Regridding / clustering

31

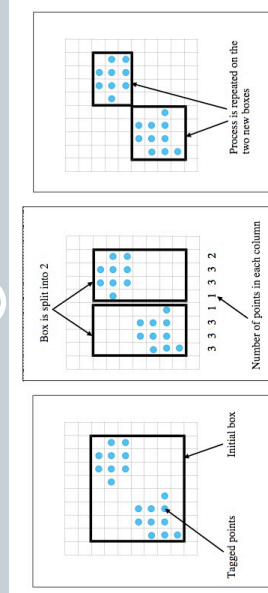


Figure 2: The 3 basic steps in regridding are (1) flag error cells and enclose in a box, (2) split the box into 2 based on a histogram of the column or row sums of tagged cells, (3) fit new boxes to each split box and repeat if the ratio of tagged to untagged cells is too small.

“Adaptive mesh refinement routines for Overture” - William Henshaw, 2011 ([link](#))

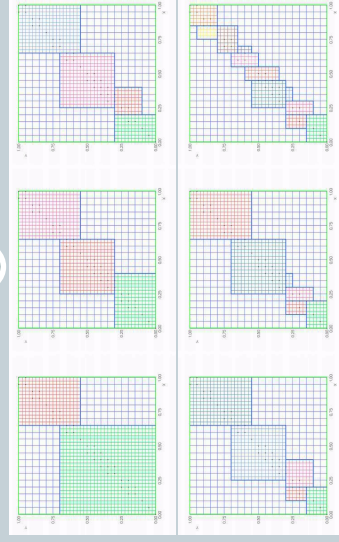
Optimal grid size, number, locations

ATMS 502 - Spring 2019

2/26/19

Regridding / clustering

32



“Adaptive mesh refinement routines for Overture” - William Henshaw, 2011 ([link](#))

Optimal grid size, number, locations

ATMS 502 - Spring 2019

2/26/19