

THE FINITE-DIFFERENCE TIME-DOMAIN (FDTD) METHOD – PART III

11. Yee's discrete algorithm

Maxwell's equations are discretized using central FDs. We set the magnetic loss equal to zero. Then,

$$\sigma_m = 0, \quad \mathbf{J}_m^i = 0$$

$$H_{x_i,j,k}^{n+0.5} = H_{x_i,j,k}^{n-0.5} - \frac{\Delta t}{\mu} \cdot \left[\frac{E_{z_i,j+1,k}^n - E_{z_i,j,k}^n}{\Delta y} - \frac{E_{y_i,j,k+1}^n - E_{y_i,j,k}^n}{\Delta z} \right]$$

$$H_{y_i,j,k}^{n+0.5} = H_{y_i,j,k}^{n-0.5} - \frac{\Delta t}{\mu} \cdot \left[\frac{E_{x_i,j,k+1}^n - E_{x_i,j,k}^n}{\Delta z} - \frac{E_{z_{i+1},j,k}^n - E_{z_i,j,k}^n}{\Delta x} \right]$$

$$H_{z_i,j,k}^{n+0.5} = H_{z_i,j,k}^{n-0.5} - \frac{\Delta t}{\mu} \cdot \left[\frac{E_{y_{i+1},j,k}^n - E_{y_i,j,k}^n}{\Delta x} - \frac{E_{x_{i,j+1},k}^n - E_{x_{i,j,k}^n}}{\Delta y} \right]$$

11. Yee's discrete algorithm – cont.

$$E_{xi,j,k}^{n+1} = k_E^E \cdot E_{xi,j,k}^n + k_H^E \cdot \left[\frac{H_{zi,j,k}^{n+0.5} - H_{zi,j-1,k}^{n+0.5}}{\Delta y} - \frac{H_{yi,j,k}^{n+0.5} - H_{yi,j,k-1}^{n+0.5}}{\Delta z} - J_{exi,j,k}^{i,n+0.5} \right]$$

$$E_{yi,j,k}^{n+1} = k_E^E \cdot E_{yi,j,k}^n + k_H^E \cdot \left[\frac{H_{xi,j,k}^{n+0.5} - H_{xi,j,k-1}^{n+0.5}}{\Delta z} - \frac{H_{zi,j,k}^{n+0.5} - H_{zi-1,j,k}^{n+0.5}}{\Delta x} - J_{eyi,j,k}^{i,n+0.5} \right]$$

$$E_{zi,j,k}^{n+1} = k_E^E \cdot E_{zi,j,k}^n + k_H^E \cdot \left[\frac{H_{yi,j,k}^{n+0.5} - H_{yi-1,j,k}^{n+0.5}}{\Delta x} - \frac{H_{xi,j,k}^{n+0.5} - H_{xi,j-1,k}^{n+0.5}}{\Delta y} - J_{ezi,j,k}^{i,n+0.5} \right]$$

$$k_E^E = \frac{1 - \frac{\sigma_e \Delta t}{2\varepsilon}}{1 + \frac{\sigma_e \Delta t}{2\varepsilon}} \quad k_H^E = \frac{\frac{\Delta t}{\varepsilon}}{1 + \frac{\sigma_e \Delta t}{2\varepsilon}} \quad \sigma_e \neq 0$$

11. Yee's discrete algorithm – cont.

The above coefficients are obtained by averaging the E -field, which appears in the loss term. For example,

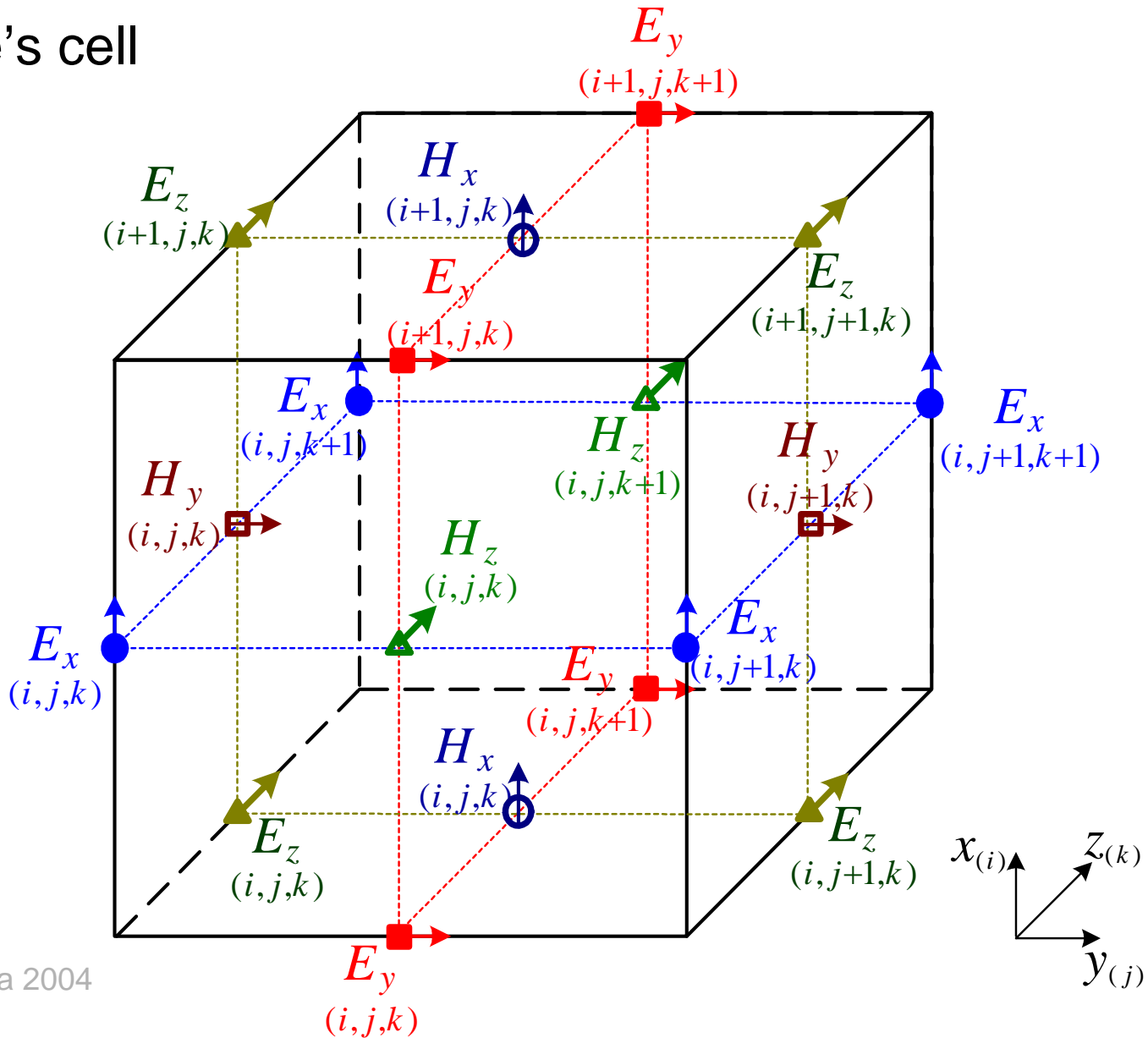
$$\varepsilon \frac{\partial E_x}{\partial t} + \sigma_e E_x = \frac{\partial H_z}{\partial y} - \frac{\partial H_y}{\partial z} - J_{ex}^i \Rightarrow$$

$$\varepsilon \frac{E_{x_{i,j,k}}^{n+1} - E_{x_{i,j,k}}^n}{\Delta t} + \sigma_e \frac{E_{x_{i,j,k}}^{n+1} + E_{x_{i,j,k}}^n}{2} =$$
$$\frac{H_{z_{i,j,k}}^{n+0.5} - H_{z_{i,j-1,k}}^{n+0.5}}{\Delta y} - \frac{H_{y_{i,j,k}}^{n+0.5} - H_{y_{i,j,k-1}}^{n+0.5}}{\Delta z} - J_{ex_{i,j,k}}^i$$

The discretization steps in time and in space, as well as the numerical constant $\alpha = c \Delta t / \Delta h$ are determined as for the wave equation.

11. Yee's discrete algorithm – cont.

Yee's cell



11. Yee's discrete algorithm – cont.

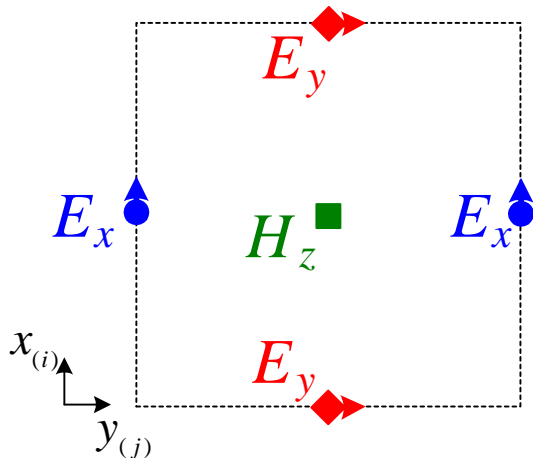
2-D problems and their discretization

The 2-D TE_z mode

$$\mu \frac{\partial H_z}{\partial t} + \sigma_m H_z = - \left(\frac{\partial E_y}{\partial x} - \frac{\partial E_x}{\partial y} \right)$$

$$\varepsilon \frac{\partial E_x}{\partial t} + \sigma_e E_x = \frac{\partial H_z}{\partial y} - J_{ex}^i$$

$$\varepsilon \frac{\partial E_y}{\partial t} + \sigma_e E_y = - \frac{\partial H_z}{\partial x} - J_{ey}^i$$

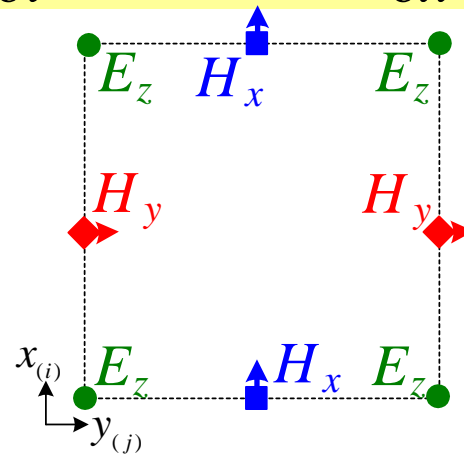


The 2-D TM_z mode

$$\varepsilon \frac{\partial E_z}{\partial t} + \sigma_e E_z = \frac{\partial H_y}{\partial x} - \frac{\partial H_x}{\partial y} - J_{ez}^i$$

$$\mu \frac{\partial H_x}{\partial t} + \sigma_m H_x = - \frac{\partial E_z}{\partial y}$$

$$\mu \frac{\partial H_y}{\partial t} + \sigma_m H_y = \frac{\partial E_z}{\partial x}$$



12. Absorbing (radiation) boundary conditions

ABCs constitute a special type of BCs, which simulate reflection-free propagation out of the computational domain. ABCs are necessary in open (radiation/scattering) problems, as well as in guided-wave problems where matched port terminations are needed.

The simplest ABCs are associated with various approximations of a one-way plane wave propagation.

- One-way wave equation (Mur's ABC)
- Liao extrapolation
- Perfectly Matched Layers – basics
- Others: Higdon operator, Bayliss-Turkel annihilating operators, etc.

12. ABCs – cont.

A. The one-way wave equation (B.Engquist and A.Majda, “Absorbing boundary conditions for the numerical simulation of waves,” *Mathematics of Computation*, vol. 31, 1977, pp. 629-651)

This is an equation which permits wave propagation in only one direction. Consider the 3-D scalar wave equation

$$\frac{\partial^2 f}{\partial x^2} + \frac{\partial^2 f}{\partial y^2} + \frac{\partial^2 f}{\partial z^2} - \frac{1}{c^2} \frac{\partial^2 f}{\partial t^2} = 0 \quad \Rightarrow \quad Lf = 0$$

The partial derivative operator is defined as

$$L = \frac{\partial^2}{\partial x^2} + \frac{\partial^2}{\partial y^2} + \frac{\partial^2}{\partial z^2} - \frac{1}{c^2} \frac{\partial^2}{\partial t^2} = \partial_x^2 + \partial_y^2 + \partial_z^2 - c^{-2} \partial_t^2$$

We wish to simulate one-way propagation along $-x$ at $x=0$.

12. ABCs – cont.

The partial differential operator L can be factored, i.e., represented as sequentially applied two operators:

$$Lf = L^+ L^- f = 0$$

$$L^- = \partial_x - c^{-1} \partial_t \sqrt{1 - S^2}$$
$$L^+ = \partial_x + c^{-1} \partial_t \sqrt{1 - S^2}$$
$$S^2 = \left(\frac{\partial_y}{c^{-1} \partial_t} \right)^2 + \left(\frac{\partial_z}{c^{-1} \partial_t} \right)^2$$

The partial differential operators L^+ and L^- are pseudo-differential operators. They cannot be applied directly to a function. Formally, the equation

$$L^- f = 0$$

represents a wave traveling along $-x$, while the equation

$$L^+ f = 0$$

represents a wave traveling along $+x$.

12. ABCs – cont.

This becomes obvious in the case of a plane wave propagating along +/- x.

$$\partial_z = 0, \partial_y = 0 \Rightarrow S = 0 \quad \Rightarrow L^- = \partial_x - c^{-1}\partial_t, \quad L^+ = \partial_x + c^{-1}\partial_t$$

$$L^- f = \frac{\partial f}{\partial x} - \frac{1}{c} \frac{\partial f}{\partial t} = 0$$

Solution: $f(x + ct)$

$$L^+ f = \frac{\partial f}{\partial x} + \frac{1}{c} \frac{\partial f}{\partial t} = 0$$

Solution: $f(x - ct)$

The radical appearing in L^+ and L^- can be expanded using Taylor series

$$(1 - S^2)^{1/2} = 1 - \frac{1}{2} S^2 + O(S^4)$$

12. ABCs – cont.

If S^2 is very small, then $(1 - S^2)^{1/2} \approx 1$

The above is a **first-order approximation of S** . This means that the partial derivatives with respect to y and z are very small when compared with the partial derivative with respect to time scaled by the velocity of propagation c .

$$S^2 = \left(\frac{\partial_y}{c^{-1}\partial_t} \right)^2 + \left(\frac{\partial_z}{c^{-1}\partial_t} \right)^2$$

This happens when the wave is incident upon the $x=\text{const.}$ plane almost normally. The L^- operator then becomes

$$\begin{aligned} L^- &= \partial_x - c^{-1}\partial_t \\ \Rightarrow L^- f &= \partial_x f - c^{-1}\partial_t f = 0 \end{aligned}$$

12. ABCs – cont.

When the wave, however, impinges upon the $x=0$ boundary wall at larger angles, the 1st order approximation is very inaccurate. At grazing angles, S is large! For better accuracy, the **second-order approximation** can be used

$$(1 - S^2)^{1/2} \approx 1 - \frac{1}{2} S^2 \quad \longleftarrow \quad S^2 = \left(\frac{\partial_y}{c^{-1} \partial_t} \right)^2 + \left(\frac{\partial_z}{c^{-1} \partial_t} \right)^2$$

The L^- operator now becomes $L^- = \partial_x - c^{-1} \partial_t \left(1 - \frac{1}{2} S^2 \right)$

$$\Rightarrow L^- = \partial_x - c^{-1} \partial_t \left[1 - \frac{1}{2} \left(\frac{\partial_y}{c^{-1} \partial_t} \right)^2 - \frac{1}{2} \left(\frac{\partial_z}{c^{-1} \partial_t} \right)^2 \right]$$

$$L^- = \partial_x - \frac{\partial_t}{c} + \frac{1}{2} \frac{c}{\partial_t} (\partial_y^2 + \partial_z^2)$$

12. ABCs – cont.

$$L^- f = \partial_{xt}^2 f - \frac{1}{c} \partial_{tt}^2 f + \frac{c}{2} (\partial_{yy}^2 f + \partial_{zz}^2 f) = 0 \quad \text{at } x = 0$$

$$L^+ f = \partial_{xt}^2 f + \frac{1}{c} \partial_{tt}^2 f - \frac{c}{2} (\partial_{yy}^2 f + \partial_{zz}^2 f) = 0 \quad \text{at } x = x_{\max}$$

$$L^- f = \partial_{yt}^2 f - \frac{1}{c} \partial_{tt}^2 f + \frac{c}{2} (\partial_{xx}^2 f + \partial_{zz}^2 f) = 0 \quad \text{at } y = 0$$

$$L^+ f = \partial_{yt}^2 f + \frac{1}{c} \partial_{tt}^2 f - \frac{c}{2} (\partial_{xx}^2 f + \partial_{zz}^2 f) = 0 \quad \text{at } y = y_{\max}$$

etc.

12. ABCs – cont.

Mur's ABC of 2nd order (G. Mur, "Absorbing boundary conditions for the finite-difference approximation of the time-domain electromagnetic field equations," *IEEE Trans. Electromagnetic Compatibility*, vol. 23, 1981, pp. 377-382.

Mur implemented the above approximate expressions into finite-difference equations. Mur expands the partial derivatives in the L^+ and L^- operators using central finite differences of the field component about an auxiliary grid point displaced half a step along the direction of absorption and along time.

Consider propagation along $-x$, at the $x=0$ boundary. We assume that the scalar function f is evaluated at integer spatial grid positions (i,j,k) and time positions n .

$$\partial_{xt}^2 f \Big|_{1/2,j,k}^n = \frac{1}{2\Delta t} \left(\frac{f_{1,j,k}^{n+1} - f_{0,j,k}^{n+1}}{\Delta x} - \frac{f_{1,j,k}^{n-1} - f_{0,j,k}^{n-1}}{\Delta x} \right)$$

12. ABCs – cont.

Now, the 2nd order time derivative has to be evaluated ½ step from the boundary as well. Mur averages the time derivatives at $x=0$ and $x=1$.

$$\partial_t^2 f = \frac{1}{2} \left[\frac{f_{0,j,k}^{n+1} - 2f_{0,j,k}^n + f_{0,j,k}^{n-1}}{\Delta t^2} + \frac{f_{1,j,k}^{n+1} - 2f_{1,j,k}^n + f_{1,j,k}^{n-1}}{\Delta t^2} \right]$$

The 2nd order y - and z - derivatives also have to be evaluated ½ step from the boundary. Mur averages those as well.

$$\partial_y^2 f = \frac{1}{2} \left[\frac{f_{0,j-1,k}^n - 2f_{0,j,k}^n + f_{0,j+1,k}^n}{\Delta y^2} + \frac{f_{1,j-1,k}^n - 2f_{1,j,k}^n + f_{1,j+1,k}^n}{\Delta y^2} \right]$$
$$\partial_z^2 f = \frac{1}{2} \left[\frac{f_{0,j,k-1}^n - 2f_{0,j,k}^n + f_{0,j,k+1}^n}{\Delta z^2} + \frac{f_{1,j,k-1}^n - 2f_{1,j,k}^n + f_{1,j,k+1}^n}{\Delta z^2} \right]$$

12. ABCs – cont.

Substitute all the FD approximations above in

$$L^- f = \partial_{xt}^2 f - \frac{1}{c} \partial_{tt}^2 f + \frac{c}{2} (\partial_{yy}^2 f + \partial_{zz}^2 f) = 0$$

The result is

$$\begin{aligned} f_{0,j,k}^{n+1} = & -f_{0,j,k}^{n-1} + k_1 (f_{1,j,k}^{n+1} + f_{0,j,k}^{n-1}) + k_2 (f_{0,j,k}^n + f_{1,j,k}^n) + \\ & + k_{3y} (f_{0,j-1,k}^n - 2f_{0,j,k}^n + f_{0,j+1,k}^n + f_{1,j-1,k}^n - 2f_{1,j,k}^n + f_{1,j+1,k}^n) + \\ & + k_{3z} (f_{0,j,k-1}^n - 2f_{0,j,k}^n + f_{0,j,k+1}^n + f_{1,j,k-1}^n - 2f_{1,j,k}^n + f_{1,j,k+1}^n) \end{aligned}$$

$$k_1 = \frac{c\Delta t - \Delta x}{c\Delta t + \Delta x}$$

$$k_2 = \frac{2\Delta x}{c\Delta t + \Delta x}$$

$$k_{3y} = \frac{(c\Delta t)^2 \Delta x}{2\Delta y^2 (c\Delta t + \Delta x)}$$

$$k_{3z} = \frac{(c\Delta t)^2 \Delta x}{2\Delta z^2 (c\Delta t + \Delta x)}$$

12. ABCs – cont.

Mur's ABC of 1st order

To obtain Mur's approximation of

$$L^- f = \partial_x f - c^{-1} \partial_t f = 0$$

simply remove the 2nd order y - and z -derivatives from the formula above:

$$f_{0,j,k}^{n+1} = -f_{0,j,k}^{n-1} + k_1 \left(f_{1,j,k}^{n+1} + f_{0,j,k}^{n-1} \right) + k_2 \left(f_{0,j,k}^n + f_{1,j,k}^n \right)$$

$$k_1 = \frac{c\Delta t - \Delta x}{c\Delta t + \Delta x}$$

$$k_2 = \frac{2\Delta x}{c\Delta t + \Delta x}$$

In Yee's algorithm, the E -field components tangential to the boundary are evaluated at this boundary. For example, at an $x=0$ boundary wall, the E_y and E_z field components define the boundary values of the EM field problem. Mur's ABC is applied to them.

12. ABCs – cont.

B. Liao's extrapolation (Z.P. Liao, H.L. Wong, B.P. Yang, and Y.F. Yuan, "A transmitting boundary for transient wave analyses," Scientia Sinica (series A), vol. XXVII, 1984, pp. 1063-1076.)

The ABC known as Liao's ABC is easily explained as an extrapolation of the wave in space-time using Newton's backward-difference polynomial. It is an order less reflective than Mur's 2nd order ABC and does not depend strongly on the angle of incidence.

We now consider a boundary wall at x_{\max} . We assume that the field values are known for points located along a straight line perpendicular to the boundary. The objective is to find an approximation of the field at the boundary at the next time step $f(x_{\max}, t + \Delta t)$.

12. ABCs – cont.

The field values used for the approximation are obtained by a simultaneous shift in space-time:

$$m = 1 \quad f_1 = f(x_{\max} - \alpha c \Delta t, t)$$

$$m = 2 \quad f_2 = f(x_{\max} - 2\alpha c \Delta t, t - \Delta t)$$

$$m = 3 \quad f_3 = f(x_{\max} - 3\alpha c \Delta t, t - 2\Delta t)$$

⋮

$$m = N \quad f_N = f(x_{\max} - N\alpha c \Delta t, t - (N - 1)\Delta t)$$

Notice that such representation corresponds to a wave propagating in the +x direction: $f(x - ct)$

We aim at finding $f_0 = f(x_{\max}, t + \Delta t)$

12. ABCs – cont.

We now define backward finite-difference approximation of p^{th} order at the point $\xi_1 = (x_{\max} - \alpha c \Delta t, t)$.

$$D^1 f(\xi_1) \equiv \Delta^1 f_1 = f_1 - f_2$$

$$D^2 f(\xi_1) \equiv \Delta^2 f_1 = \Delta^1 f_1 - \Delta^1 f_2, \quad \Delta^1 f_2 = f_2 - f_3$$

$$D^3 f(\xi_1) \equiv \Delta^3 f_1 = \Delta^2 f_1 - \Delta^2 f_2, \quad \Delta^2 f_2 = \Delta^1 f_2 - \Delta^1 f_3$$

$$\vdots \quad \Delta^1 f_3 = f_3 - f_4$$

$$D^N f(\xi_1) \equiv \Delta^N f_1 = \Delta^{N-1} f_1 - \Delta^{N-1} f_2$$

We denote the discrete functions, which go back in space and time as (see previous slide)

$$f_m = f(\xi_m) = f(x_{\max} - m\alpha c \Delta t, t - (m-1)\Delta t)$$

12. ABCs – cont.

The N -th backward difference can be written in terms of the function values f_m as

$$\Delta^N f_1 = \sum_{m=1}^{N+1} (-1)^{m-1} C_{m-1}^N f_m,$$

where the Newton binomial coefficients are

$$C_{m-1}^N = \binom{N}{m-1} = \frac{N!}{(N-m+1)!(m-1)!}$$

However, it is usually more efficient (and easy) to program the recursive calculations shown in the previous slide.

12. ABCs – cont.

We can express (interpolate) the function in terms of the backward finite differences at f_1 as

$$f_m \cong f_1 + \beta \Delta^1 f_1 + \frac{\beta(\beta+1)}{2!} \Delta^2 f_1 + \frac{\beta(\beta+1)(\beta+2)}{3!} \Delta^3 f_1 + \dots \\ + \frac{\beta(\beta+1)\dots(\beta+N-2)}{(N-1)!} \Delta^{N-1} f_1, \quad 1 \leq m \leq N, \quad \beta = 1 - m$$

We now use the above formula to **extrapolate** the function values, and we set $m = 0$. Then, $\beta = 1$.

$$f_0 = f(t + \Delta t, x_{\max}) \cong f_1 + \Delta^1 f_1 + \Delta^2 f_1 + \Delta^3 f_1 + \dots + \Delta^{N-1} f_1$$

This is Liao's ABC.

12. ABCs – cont.

Liao *et al.* showed that for a sinusoidal plane wave of unit amplitude and wavelength λ , the maximum error is given by

$$|\Delta^N f|_{\max} = 2^N \sin^N(\pi c \Delta t / \lambda)$$

Assuming that

$$\Delta h = 2c \Delta t \quad \text{and} \quad \Delta h = \lambda / 32$$

the error is estimated at 0.1%. Liao's ABC is robust and depends little on the angle of incidence. With orders higher than $N=3$, however, it sometimes causes instabilities.

12. ABCs – cont.

$$f_0 = f(t + \Delta t, x_{\max}) \cong f_1 + \Delta^1 f_1 + \Delta^2 f_1 + \Delta^3 f_1 + \cdots \Delta^{N-1} f_1$$

Liao's ABC is simple to implement.

! ABC in MAIN

ix2=nt-((nt-1)/2)*2

ix3=nt-((nt-1)/3)*3

ix4=nt-((nt-1)/4)*4

F0=>AX(:, :, nk); F1=>AX(:, :, nk-1) ! front, ABC
F2=>A2_F(:, :, ix2); F3=>A3_F(:, :, ix3); F4=>A4_F(:, :, ix4)
call LIAO(ni, nj)

! before end of cycle
! history
A2_F(:, :, ix2)=AX(:, :, 3)
A3_F(:, :, ix3)=AX(:, :, 4)
A4_F(:, :, ix4)=AX(:, :, 5)

subroutine LIAO

D1=F1-F2; D2=F2-F3; D3=F3-F4

DD1=D1-D2; DD2=D2-D3

DDD1=DD1-DD2

F0=F1+D1+DD1+DDD1

return

end subroutine LIAO

! define variables
real(8), dimension(ni, nj, 2), target:: A2_F, A2_B
real(8), dimension(ni, nj, 3), target:: A3_F, A3_B
real(8), dimension(ni, nj, 4), target:: A4_F, A4_B
real(8), dimension(:, :), pointer:: F0, F1, F2, F3, F4