

FDTD: An Introduction

- PML could be analyzed in stretched coordinate systems
- Source-free Maxwell's equations

$$\nabla_{\text{stretched}} \times \vec{E} = -j\omega\mu\vec{H}; \nabla_{\text{stretched}} \times \vec{H} = j\omega\epsilon\vec{E};$$

$$\nabla_{\text{stretched}} \bullet (\epsilon \vec{E}) = 0; \nabla_{\text{stretched}} \bullet (\mu \vec{H}) = 0;$$

$$\nabla_{\text{stretched}} = \hat{x} \frac{1}{s_x(x)} \frac{\partial}{\partial x} + \hat{y} \frac{1}{s_y(y)} \frac{\partial}{\partial y} + \hat{z} \frac{1}{s_z(z)} \frac{\partial}{\partial z}$$

- Using this new definition of del operator, we can find components for the two Maxwell's curl equations

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For example

- x-component of first Maxwell's curl equation is

$$\frac{1}{s_y} \frac{\partial H_z}{\partial y} - \frac{1}{s_z} \frac{\partial H_y}{\partial z} = j\omega\epsilon E_x$$

- In time domain, multiplication becomes convolution

$$\bar{S}_y * \frac{\partial H_z}{\partial y} - \bar{S}_z * \frac{\partial H_y}{\partial z} = \epsilon \frac{\partial E_x}{\partial t}; \bar{S}_y = F^{-1} \left(\frac{1}{s_y} \right), \bar{S}_z = F^{-1} \left(\frac{1}{s_z} \right)$$

- * means convolution, that's why convolutional PML, supposedly the best PML

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- Note

$$\frac{1}{s_x} = \frac{1}{k_x + \frac{\sigma_x}{\alpha_x + j\omega\varepsilon}} = \frac{\alpha_x + j\omega\varepsilon}{\alpha_x k_x + j\omega\varepsilon k_x + \sigma_x}$$

- which can be expressed as

$$\frac{1}{s_x} = \frac{1}{k_x} - \frac{1}{B_x}; \frac{1}{B_x} = \frac{\sigma_x}{k_x} \left(\frac{1}{\alpha_x k_x + j\omega\varepsilon k_x + \sigma_x} \right)$$

$$\frac{1}{s_i} = \frac{1}{k_i} - \frac{1}{B_i}; \frac{1}{B_i} = \frac{\sigma_i}{k_i} \left(\frac{1}{\alpha_i k_i + j\omega\varepsilon k_i + \sigma_i} \right), i \in \{x, y, z\}$$

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- Therefore, inverse FT of inverse of scale factor s_i

$$\begin{aligned}\bar{S}_i &= F^{-1} \left(\frac{1}{s_i} \right) = F^{-1} \left(\frac{1}{k_i} - \frac{\sigma_i}{k_i} \left(\frac{1}{\alpha_i k_i + j\omega \epsilon k_i + \sigma_i} \right) \right) \\ &= F^{-1} \left(\frac{1}{k_i} - \frac{\sigma_i}{k_i \epsilon k_i} \left(\frac{1}{\frac{\alpha_i}{\epsilon} + j\omega + \frac{\sigma_i}{\epsilon k_i}} \right) \right) \\ &= \frac{\delta(t)}{k_i} - \frac{\sigma_i}{k_i^2 \epsilon} \exp \left(- \left(\frac{\alpha_i}{\epsilon} + \frac{\sigma_i}{k_i \epsilon} \right) t \right) = \frac{\delta(t)}{k_i} + Q_i(t)\end{aligned}$$

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- Hence,

$$\frac{1}{k_y} \frac{\partial H_z}{\partial y} - \frac{1}{k_z} \frac{\partial H_y}{\partial z} + Q_y(t)^* \frac{\partial H_z}{\partial y} - Q_z(t)^* \frac{\partial H_y}{\partial z} = \epsilon \frac{\partial E_x}{\partial t}$$

- Define a new Ψ function for the 3rd and 4th convolution terms in the above equation

$$\Psi_{E_j w}^n = Q_w(t)^* \frac{\partial H_k}{\partial w}; t = n\Delta t, E_j \in (E_x, E_y, E_z), w, k \in (x, y, z); j \neq w \neq k$$

- Notation for $\Psi_{E_j w}^n$

- Subscript of Ψ reveals that this function appears in the update of E_j
- Subscript w shows that spatial derivative is with respect to w
- Superscript n is the time running index

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$$\Psi_{E_j w}^n = Q_w(t)^* \frac{\partial H_k}{\partial w}; t = n\Delta t, E_j \in (E_x, E_y, E_z), w, k \in (x, y, z); j \neq w \neq k$$

- Note that in this integration

$$\Psi_{E_j w}^n = \int_{\tau=0}^{n\Delta t} Q_w(\tau) \frac{\partial H_k(n\Delta t - \tau)}{\partial w} d\tau$$

- The integrand is zero for $\tau < 0$ since $Q_w(\tau) = 0, \tau < 0$
- and $\tau = 0$ is lower limit of integration
- Also the upper limit on the integration $\tau = n\Delta t$ can be observed from

$$H_k(n\Delta t - \tau) = 0; n\Delta t - \tau < 0 \Rightarrow H_k(n\Delta t - \tau) = 0; n\Delta t < \tau$$

- Second function and integrand is zero for $n\Delta t < \tau$

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- In FDTD, H_k varies in discrete time

$$\frac{\partial H_k(t)}{\partial w} = \sum_{j=0}^{J_{\max}} \frac{\partial H_k(n\Delta t)}{\partial w} p_j(t); p_j(t) = \begin{cases} 1, & j\Delta t \leq t < (j+1)\Delta t \\ 0, & \text{otherwise} \end{cases}$$

To find $\frac{\partial H_k(n\Delta t - \tau)}{\partial w}$

Note that $H_k(n\Delta t - \tau)$ is basically flipping and shifting

Hence

$$f(n\Delta t - \tau) = \sum_{j=0}^{n-1} f_{n-j} p_j(\tau)$$

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$$\Rightarrow \frac{\partial H_k(n\Delta t - \tau)}{\partial w} = \sum_{j=0}^{n-1} \frac{\partial H_k^{n-j}}{\partial w} p_j(\tau)$$

At time step n



$$\Psi_{E_j w}^n = \int_{\tau=0}^{n\Delta t} Q_w(\tau) \sum_{j=0}^{n-1} \frac{\partial H_k^{n-j}}{\partial w} p_j(\tau) d\tau$$

Interchanging the summation and integration

$$\Psi_{E_j w}^n = \sum_{j=0}^{n-1} \frac{\partial H_k^{n-j}}{\partial w} \int_{\tau=0}^{n\Delta t} Q_w(\tau) p_j(\tau) d\tau$$

Pulse function is equal to 1 for the interval $j\Delta t \leq t < (j+1)\Delta t$

$$\Psi_{E_j w}^n = \sum_{j=0}^{n-1} \frac{\partial H_k^{n-j}}{\partial w} \int_{\tau=j\Delta t}^{\tau=(j+1)\Delta t} Q_w(\tau) d\tau$$

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Since Q_w is an exponential function, we can easily find it's integration as

$$\int_{\tau=j\Delta t}^{\tau=(j+1)\Delta t} Q_w(\tau) d\tau = \int_{\tau=j\Delta t}^{\tau=(j+1)\Delta t} \frac{\sigma_i}{k_i^2 \epsilon} \exp\left(-\left(\frac{\alpha_i}{\epsilon} + \frac{\sigma_i}{k_i \epsilon}\right)t\right) d\tau = C_w (b_w)^j$$

- where

$$b_w = \exp\left(-\left(\frac{\alpha_i}{\epsilon} + \frac{\sigma_i}{k_i \epsilon}\right)\Delta t\right); C_w = \frac{\sigma_w}{\sigma_w k_w + (k_w)^2 \alpha_w} (b_w - 1)$$

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- Now $\Psi_{E_j w}^n = \sum_{j=0}^{n-1} \frac{\partial H_k^{n-j}}{\partial w} C_w (b_w)^j$
- Separating $j=0$ and $j=1, \dots, n-1$ terms, we have,

$$\Psi_{E_j w}^n = C_w \frac{\partial H_k^n}{\partial w} + \sum_{j=1}^{n-1} \frac{\partial H_k^{n-j}}{\partial w} C_w (b_w)^j$$

- Putting $j=l+1$

$$\Psi_{E_j w}^n = C_w \frac{\partial H_k^n}{\partial w} + \sum_{l=0}^{n-2} \frac{\partial H_k^{n-l-1}}{\partial w} C_w (b_w)^{l+1}$$

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- Taking out b_w as a common factor in the second term

$$\Psi_{E_j w}^n = C_w \frac{\partial H_k^n}{\partial w} + b_w \left\{ \sum_{l=0}^{(n-1)-1} \frac{\partial H_k^{(n-1)-l}}{\partial w} C_w (b_w)^l \right\}$$

- Observe that the term in second bracket in the above expression is exactly $\Psi_{E_j w}^{n-1}$
- Hence, the update equation for

$$\Psi_{E_j w}^n = C_w \frac{\partial H_k^n}{\partial w} + b_w \Psi_{E_j w}^{n-1}$$

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$$\Psi_{E_j w}^n = C_w \frac{\partial H_k^n}{\partial w} + b_w \Psi_{E_j w}^{n-1}$$

- Discretization of CPML

$$\epsilon \frac{\partial E_x}{\partial t} = \frac{1}{k_y} \frac{\partial H_z}{\partial y} - \frac{1}{k_z} \frac{\partial H_y}{\partial z} + \boxed{Q_y(t)^* \frac{\partial H_z}{\partial y}} - \boxed{Q_z(t)^* \frac{\partial H_y}{\partial z}}$$

$$\begin{aligned} & \epsilon \frac{E_x|_{i+1/2,j,k}^{n+1} - E_x|_{i+1/2,j,k}^n}{\Delta t} \\ &= \frac{H_z|_{i+1/2,j+1/2,k}^{n+1/2} - H_z|_{i+1/2,j-1/2,k}^{n+1/2}}{k_y \Delta y} - \frac{H_y|_{i+1/2,j,k+1/2}^{n+1/2} - H_y|_{i+1/2,j,k-1/2}^{n+1/2}}{k_z \Delta z} \end{aligned}$$

$$+ \Psi_{E_x,y}|_{i+1/2,j,k}^{n+1/2} - \Psi_{E_x,z}|_{i+1/2,j,k}^{n+1/2}$$

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- and

$$\Psi_{E_j w}^n = b_w \Psi_{E_j w}^{n-1} + C_w \frac{\partial H_k^n}{\partial w}$$

$$\Psi_{E_x, y} \Big|_{i+1/2, j, k}^{n+1/2} = b_y \Psi_{E_x, y} \Big|_{i+1/2, j, k}^{n-1/2} + C_y \frac{H_z \Big|_{i+1/2, j+1/2, k}^{n+1/2} - H_z \Big|_{i+1/2, j-1/2, k}^{n+1/2}}{\Delta y}$$

$$\Psi_{E_x, z} \Big|_{i+1/2, j, k}^{n+1/2} = b_z \Psi_{E_x, z} \Big|_{i+1/2, j, k}^{n-1/2} + C_z \frac{H_y \Big|_{i+1/2, j, k+1/2}^{n+1/2} - H_y \Big|_{i+1/2, j, k-1/2}^{n+1/2}}{\Delta z}$$

- where

$$b_w = \exp \left(- \left(\frac{\alpha_i}{\epsilon} + \frac{\sigma_i}{k_i \epsilon} \right) \Delta t \right); C_w = \frac{\sigma_w}{\sigma_w k_w + (k_w)^2 \alpha_w} (b_w - 1)$$

FDTD: Advances

- Unconditional stable FDTD
 - Time step and space step can be chosen independently
 - For example, Crank Nicolson (CN) FDTD
 - Time step size can be chosen beyond CFL limit
- CN FDTD
 - Time and space derivatives are discretized by centered differences
 - Fields affected by the curl operators are averaged in time

FDTD: Advances

$$\frac{\partial E_x}{\partial t} = \frac{1}{\epsilon_0 \epsilon_r} \left(\frac{\partial H_z}{\partial y} - \frac{\partial H_y}{\partial z} \right)$$

$$\frac{\partial E_y}{\partial t} = \frac{1}{\epsilon_0 \epsilon_r} \left(\frac{\partial H_x}{\partial z} - \frac{\partial H_z}{\partial x} \right)$$

$$\frac{\partial E_z}{\partial t} = \frac{1}{\epsilon_0 \epsilon_r} \left(\frac{\partial H_y}{\partial x} - \frac{\partial H_x}{\partial y} \right)$$

$$\frac{\partial H_x}{\partial t} = \frac{1}{\mu_0 \mu_r} \left(\frac{\partial E_y}{\partial z} - \frac{\partial E_z}{\partial y} \right)$$

$$\frac{\partial H_y}{\partial t} = \frac{1}{\mu_0 \mu_r} \left(\frac{\partial E_z}{\partial x} - \frac{\partial E_x}{\partial z} \right)$$

$$\frac{\partial H_z}{\partial t} = \frac{1}{\mu_0 \mu_r} \left(\frac{\partial E_x}{\partial y} - \frac{\partial E_y}{\partial x} \right)$$

- Can convert these Maxwell's equations into matrix equation as

$$\frac{\partial \mathbf{U}}{\partial t} = \mathbf{A}\mathbf{U} + \mathbf{B}\mathbf{U} = (\mathbf{A} + \mathbf{B})\mathbf{U}$$

FDTD: Advances

- where

$$\mathbf{U} = \begin{bmatrix} E_x \\ E_y \\ E_z \\ H_x \\ H_y \\ H_z \end{bmatrix} \quad \mathbf{A} = \begin{bmatrix} 0 & 0 & 0 & 0 & 0 & \frac{1}{\varepsilon} \frac{\partial}{\partial y} \\ 0 & 0 & 0 & \frac{1}{\varepsilon} \frac{\partial}{\partial z} & 0 & 0 \\ 0 & 0 & 0 & 0 & \frac{1}{\varepsilon} \frac{\partial}{\partial x} & 0 \\ 0 & \frac{1}{\mu} \frac{\partial}{\partial z} & 0 & 0 & 0 & 0 \\ 0 & 0 & \frac{1}{\mu} \frac{\partial}{\partial x} & 0 & 0 & 0 \\ \frac{1}{\mu} \frac{\partial}{\partial y} & 0 & 0 & 0 & 0 & 0 \end{bmatrix}$$

FDTD: Advances

- and

$$\mathbf{B} = \begin{bmatrix} 0 & 0 & 0 & 0 & -\frac{1}{\epsilon} \frac{\partial}{\partial z} & 0 \\ 0 & 0 & 0 & 0 & 0 & -\frac{1}{\epsilon} \frac{\partial}{\partial x} \\ 0 & 0 & 0 & -\frac{1}{\epsilon} \frac{\partial}{\partial y} & 0 & 0 \\ 0 & 0 & -\frac{1}{\mu} \frac{\partial}{\partial y} & 0 & 0 & 0 \\ -\frac{1}{\mu} \frac{\partial}{\partial z} & 0 & 0 & 0 & 0 & 0 \\ 0 & -\frac{1}{\mu} \frac{\partial}{\partial x} & 0 & 0 & 0 & 0 \end{bmatrix}$$

FDTD: Advances

- Time and space derivatives are discretized by centered differences
- Fields affected by the curl operators are averaged in time

$$\frac{\partial \mathbf{U}}{\partial t} = \mathbf{A}\mathbf{U} + \mathbf{B}\mathbf{U} = (\mathbf{A} + \mathbf{B})\mathbf{U}$$

$$\frac{\mathbf{U}^{n+1} - \mathbf{U}^n}{\Delta t} = (\mathbf{A} + \mathbf{B}) \frac{\mathbf{U}^{n+1} + \mathbf{U}^n}{2}$$

FDTD: Advances

- Rearranging,

$$\therefore \mathbf{U}^{n+1} - \mathbf{U}^n = \left(\frac{\Delta t}{2} \mathbf{A} + \frac{\Delta t}{2} \mathbf{B} \right) (\mathbf{U}^{n+1} + \mathbf{U}^n)$$

- Combining terms,

$$\Rightarrow \mathbf{U}^{n+1} - \left(\frac{\Delta t}{2} \mathbf{A} + \frac{\Delta t}{2} \mathbf{B} \right) \mathbf{U}^{n+1} = \mathbf{U}^n + \left(\frac{\Delta t}{2} \mathbf{A} + \frac{\Delta t}{2} \mathbf{B} \right) \mathbf{U}^n$$

$$\Rightarrow \left(\mathbf{I} - \frac{\Delta t}{2} \mathbf{A} - \frac{\Delta t}{2} \mathbf{B} \right) \mathbf{U}^{n+1} = \left(\mathbf{I} + \frac{\Delta t}{2} \mathbf{A} + \frac{\Delta t}{2} \mathbf{B} \right) \mathbf{U}^n$$

$$\because (1 \pm a_1)(1 \pm a_2) = 1 \pm a_1 \pm a_2 \pm a_1 a_2$$

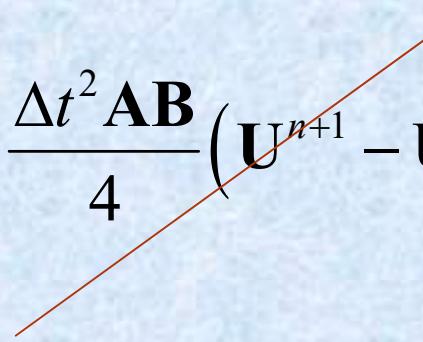
FDTD: Advances

So adding $\frac{\Delta t^2 \mathbf{AB}}{4}$ on both sides of the above equation, we have,

$$\begin{aligned} & \left(\mathbf{I} - \frac{\Delta t}{2} \mathbf{A} - \frac{\Delta t}{2} \mathbf{B} + \frac{\Delta t^2 \mathbf{AB}}{4} \right) \mathbf{U}^{n+1} \\ &= \left(\mathbf{I} + \frac{\Delta t}{2} \mathbf{A} + \frac{\Delta t}{2} \mathbf{B} + \frac{\Delta t^2 \mathbf{AB}}{4} \right) \mathbf{U}^n + \frac{\Delta t^2 \mathbf{AB}}{4} (\mathbf{U}^{n+1} - \mathbf{U}^n) \\ &\Rightarrow \left(\mathbf{I} - \frac{\Delta t}{2} \mathbf{A} \right) \left(\mathbf{I} - \frac{\Delta t}{2} \mathbf{B} \right) \mathbf{U}^{n+1} \\ &= \left(\mathbf{I} + \frac{\Delta t}{2} \mathbf{A} \right) \left(\mathbf{I} + \frac{\Delta t}{2} \mathbf{B} \right) \mathbf{U}^n + \frac{\Delta t^2 \mathbf{AB}}{4} (\mathbf{U}^{n+1} - \mathbf{U}^n) \end{aligned}$$

FDTD: Advances

- Approximate implementation
- Alternating direction implicit (ADI-FDTD) method
- It drops the last term in the equation

$$\begin{aligned} & \left(\mathbf{I} - \frac{\Delta t}{2} \mathbf{A} \right) \left(\mathbf{I} - \frac{\Delta t}{2} \mathbf{B} \right) \mathbf{U}^{n+1} \\ &= \left(\mathbf{I} + \frac{\Delta t}{2} \mathbf{A} \right) \left(\mathbf{I} + \frac{\Delta t}{2} \mathbf{B} \right) \mathbf{U}^n + \frac{\Delta t^2 \mathbf{AB}}{4} (\mathbf{U}^{n+1} - \mathbf{U}^n) \end{aligned}$$


FDTD: Advances

- And split it into two steps
- Step 1:

$$\left(\mathbf{I} - \frac{\Delta t}{2} \mathbf{A} \right) \mathbf{U}^{n+\frac{1}{2}} = \left(\mathbf{I} + \frac{\Delta t}{2} \mathbf{B} \right) \mathbf{U}^n$$

- Step 2:

$$\left(\mathbf{I} - \frac{\Delta t}{2} \mathbf{B} \right) \mathbf{U}^{n+1} = \left(\mathbf{I} + \frac{\Delta t}{2} \mathbf{A} \right) \mathbf{U}^{n+\frac{1}{2}}$$

FDTD: Advances

- In ADI-FDTD, computational overhead is smaller
- Because of omission of the last term,
 - the ADI-FDTD method leads to truncation error
 - which is a function of $\frac{\Delta t^2}{4}$ times the space derivatives of the fields
- Hence imposing a limit on Δt
- ADI-FDTD improves the computational efficiency at the cost of accuracy

FDTD: Advances

- Alternating direction implicit (ADI) FDTD

$$\begin{aligned}\frac{\partial E_x}{\partial t} &= \frac{1}{\epsilon_0 \epsilon_r} \left(\frac{\partial H_z}{\partial y} - \frac{\partial H_y}{\partial z} \right) & \frac{\partial H_x}{\partial t} &= \frac{1}{\mu_0 \mu_r} \left(\frac{\partial E_y}{\partial z} - \frac{\partial E_z}{\partial y} \right) \\ \frac{\partial E_y}{\partial t} &= \frac{1}{\epsilon_0 \epsilon_r} \left(\frac{\partial H_x}{\partial z} - \frac{\partial H_z}{\partial x} \right) & \frac{\partial H_y}{\partial t} &= \frac{1}{\mu_0 \mu_r} \left(\frac{\partial E_z}{\partial x} - \frac{\partial E_x}{\partial z} \right) \\ \frac{\partial E_z}{\partial t} &= \frac{1}{\epsilon_0 \epsilon_r} \left(\frac{\partial H_y}{\partial x} - \frac{\partial H_x}{\partial y} \right) & \frac{\partial H_z}{\partial t} &= \frac{1}{\mu_0 \mu_r} \left(\frac{\partial E_x}{\partial y} - \frac{\partial E_y}{\partial x} \right)\end{aligned}$$

- Two-step scheme in time
 - First half time step (First procedure)
 - Second half time step (Second procedure)

FDTD: Advances

- 3 equations for 2-D TM^z are as follows

$$\frac{\partial E_z}{\partial t} = \frac{1}{\epsilon} \left(\frac{\partial H_y}{\partial x} - \frac{\partial H_x}{\partial y} \right)$$

$$\frac{\partial H_x}{\partial t} = -\frac{1}{\mu} \left(\frac{\partial E_z}{\partial y} \right)$$

$$\frac{\partial H_y}{\partial t} = \frac{1}{\mu} \left(\frac{\partial E_z}{\partial x} \right)$$

FDTD: Advances

- First procedure:

$$E_z|_{i,j}^{n+1/2} = E_z|_{i,j}^n$$

$$+ \frac{\Delta t}{2\epsilon\Delta x} \left(H_y|_{i+1/2,j}^{n+1/2} - H_y|_{i-1/2,j}^{n+1/2} \right) - \frac{\Delta t}{2\epsilon\Delta y} \left(H_x|_{i,j+1/2}^n - H_x|_{i,j-1/2}^n \right)$$



$$\frac{\partial E_z}{\partial t} = \frac{1}{\epsilon} \left(\frac{\partial H_y}{\partial x} - \frac{\partial H_x}{\partial y} \right)$$

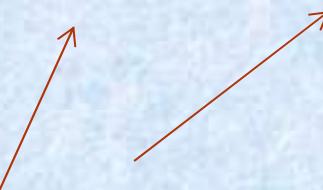
- Implicit:
 - How do you calculate?

FDTD: Advances

- We can use 3rd equation

$$\frac{\partial H_y}{\partial t} = \frac{1}{\mu} \left(\frac{\partial E_z}{\partial x} \right)$$

$$H_y|_{i+1/2,j}^{n+1/2} = H_y|_{i+1/2,j}^n - \frac{\Delta t}{2\mu\Delta x} \left(E_z|_{i+1,j}^{n+1/2} - E_z|_{i,j}^{n+1/2} \right)$$



FDTD: Advances

- Substituting equation (3) in equation (1), we have,

$$\begin{aligned} E_z|_{i,j}^{n+1/2} &= E_z|_{i,j}^n \\ &+ \frac{\Delta t}{2\epsilon\Delta x} \left(H_y|_{i+1/2,j}^n - \frac{\Delta t}{2\mu\Delta x} \left(E_z|_{i+1,j}^{n+1/2} - E_z|_{i,j}^{n+1/2} \right) - \left(H_y|_{i-1/2,j}^n - \frac{\Delta t}{2\mu\Delta x} \left(E_z|_{i,j}^{n+1/2} - E_z|_{i-1,j}^{n+1/2} \right) \right) \right) \\ &- \frac{\Delta t}{2\epsilon\Delta y} \left(H_x|_{i,j+1/2}^n - H_x|_{i,j-1/2}^n \right) \end{aligned}$$

FDTD: Advances

- Rearranging

$$\begin{aligned} & \alpha_1 E_z|_{i-1,j}^{n+1/2} + \beta_1 E_z|_{i,j}^{n+1/2} + \gamma_1 E_z|_{i+1,j}^{n+1/2} \\ &= E_z|_{i,j}^n + \frac{\Delta t}{2\epsilon\Delta x} \left(H_y|_{i+1/2,j}^n - H_y|_{i-1/2,j}^n \right) - \frac{\Delta t}{2\epsilon\Delta y} \left(H_x|_{i,j+1/2}^n - H_x|_{i,j-1/2}^n \right) \end{aligned}$$

- where

$$\alpha_1 = -\frac{\Delta t}{2\mu\Delta x} \frac{\Delta t}{2\epsilon\Delta x}; \beta_1 = 1 - \alpha_1 - \gamma_1; \gamma_1 = -\frac{\Delta t}{2\mu\Delta x} \frac{\Delta t}{2\epsilon\Delta x};$$

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- In matrix form

$$\begin{bmatrix} \beta_1(1) & \gamma_1(1) & 0 & \cdots & 0 \\ \alpha_1(2) & \beta_1(2) & \gamma_1(2) & \cdots & 0 \\ \vdots & \cdots & \ddots & \cdots & \vdots \\ 0 & \cdots & \alpha_1(m-1) & \beta_1(m-1) & \gamma_1(m-1) \\ 0 & \cdots & 0 & \alpha_1(m) & \beta_1(m) \end{bmatrix} \begin{bmatrix} E_z^{n+1/2}(j,1) \\ E_z^{n+1/2}(j,2) \\ \vdots \\ E_z^{n+1/2}(j,m-1) \\ E_z^{n+1/2}(j,m) \end{bmatrix} = \begin{bmatrix} r_1^n(1) \\ r_1^n(2) \\ \vdots \\ r_1^n(m-1) \\ r_1^n(m) \end{bmatrix}$$

FDTD: Advances

- Second procedure:

$$E_z|_{i,j}^{n+1} = E_z|_{i,j}^{n+1/2}$$

$$+ \frac{\Delta t}{2\epsilon\Delta x} \left(H_y|_{i+1/2,j}^{n+1/2} - H_y|_{i-1/2,j}^{n+1/2} \right) - \frac{\Delta t}{2\epsilon\Delta y} \left(H_x|_{i,j+1/2}^{n+1} - H_x|_{i,j-1/2}^{n+1} \right)$$

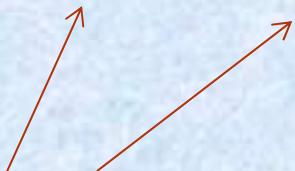
- Implicit:
 - How do you calculate?

FDTD: Advances

- Use the 2nd equation

$$\frac{\partial H_x}{\partial t} = -\frac{1}{\mu} \left(\frac{\partial E_z}{\partial y} \right)$$

$$H_x|_{i,j+1/2}^{n+1} = H_x|_{i,j+1/2}^{n+1/2} - \frac{\Delta t}{2\mu\Delta y} \left(E_z|_{i,j+1}^{n+1} - E_z|_{i,j}^{n+1} \right)$$



FDTD: Advances

- Substituting equation (2) in equation (1), we have,

$$\begin{aligned} E_z|_{i,j}^{n+1} &= E_z|_{i,j}^{n+1/2} + \frac{\Delta t}{2\epsilon\Delta x} \left(H_y|_{i+1/2,j}^{n+1/2} - H_y|_{i-1/2,j}^{n+1/2} \right) \\ &\quad - \frac{\Delta t}{2\epsilon\Delta y} \left(H_x|_{i,j+1/2}^{n+1/2} - \frac{\Delta t}{2\mu\Delta y} \left(E_z|_{i,j+1}^{n+1} - E_z|_{i,j}^{n+1} \right) - H_x|_{i,j-1/2}^{n+1/2} - \frac{\Delta t}{2\mu\Delta y} \left(E_z|_{i,j}^{n+1} - E_z|_{i,j-1}^{n+1} \right) \right) \end{aligned}$$

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- Rearranging

$$\begin{aligned} & \alpha_2 E_z|_{i,j-1}^{n+1} + \beta_2 E_z|_{i,j}^{n+1} + \gamma_2 E_z|_{i,j+1}^{n+1} \\ = & \boxed{E_z|_{i,j}^{n+1/2} + \frac{\Delta t}{2\epsilon\Delta y} \left(H_y|_{i+1/2,j}^{n+1/2} - H_y|_{i-1/2,j}^{n+1/2} \right) - \frac{\Delta t}{2\mu\Delta x} \left(H_x|_{i,j+1/2}^{n+1/2} - H_x|_{i,j-1/2}^{n+1/2} \right)} \end{aligned}$$

- where

$$\alpha_2 = -\frac{\Delta t}{2\mu\Delta x} \frac{\Delta t}{2\epsilon\Delta x}; \beta_2 = 1 - \alpha_2 - \gamma_2; \gamma_2 = -\frac{\Delta t}{2\mu\Delta x} \frac{\Delta t}{2\epsilon\Delta x};$$

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- In matrix form

$$\begin{bmatrix} \beta_1(1) & \gamma_1(1) & 0 & \cdots & 0 \\ \alpha_1(2) & \beta_1(2) & \gamma_1(2) & \cdots & 0 \\ \vdots & \cdots & \ddots & \cdots & \vdots \\ 0 & \cdots & \alpha_1(m-1) & \beta_1(m-1) & \gamma_1(m-1) \\ 0 & \cdots & 0 & \alpha_1(m) & \beta_1(m) \end{bmatrix} \begin{bmatrix} E_z^{n+1}(j,1) \\ E_z^{n+1}(j,2) \\ \vdots \\ E_z^{n+1}(j,m-1) \\ E_z^{n+1}(j,m) \end{bmatrix} = \begin{bmatrix} r_1^{n+1/2}(1) \\ r_1^{n+1/2}(2) \\ \vdots \\ r_1^{n+1/2}(m-1) \\ r_1^{n+1/2}(m) \end{bmatrix}$$

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- How do you find the inverse of a tridiagonal matrix?
- Consider a tridiagonal matrix such as

$$\begin{pmatrix} A_{11} & A_{12} & 0 & 0 \\ A_{21} & A_{22} & A_{23} & 0 \\ 0 & A_{32} & A_{33} & A_{34} \\ 0 & 0 & A_{43} & A_{44} \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \end{pmatrix} = \begin{pmatrix} b_1 \\ b_2 \\ b_3 \\ b_4 \end{pmatrix}$$

- First make it Upper triangular and do back substitution

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- Conversion from Tridiagonal matrix to Upper triangular matrix
- Multiply first equation by A_{21}/A_{11} and subtract it from second equation

$$\begin{pmatrix} A_{11} & A_{12} & 0 & 0 \\ 0 & A_{22} - \frac{A_{21}A_{12}}{A_{11}} & A_{23} & 0 \\ 0 & A_{32} & A_{33} & A_{34} \\ 0 & 0 & A_{43} & A_{44} \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \end{pmatrix} = \begin{pmatrix} b_1 \\ b_2 - \frac{A_{21}b_1}{A_{11}} \\ b_3 \\ b_4 \end{pmatrix}$$

- Continue the process, we will get an upper triangular matrix

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- Upper triangular matrix looks like this

$$\begin{pmatrix} A_{11} & A_{12} & 0 & 0 \\ 0 & A_{22} & A_{23} & 0 \\ 0 & 0 & A_{33} & A_{34} \\ 0 & 0 & 0 & A_{44} \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \end{pmatrix} = \begin{pmatrix} b_1 \\ b_2 \\ b_3 \\ b_4 \end{pmatrix}$$

- Consider the last equation

$$x_4 = \frac{b_4}{A_{44}}; A_{33}x_3 + A_{34}x_4 = b_3 \Rightarrow x_3 = \frac{b_3 - A_{34}x_4}{A_{33}}; x_2 = \frac{b_2 - A_{23}x_3}{A_{22}}, \dots$$