

# Introduction to the FDTD method

Ilkka Laakso

Department of Electrical Engineering and Automation

Tfy-99.3227

4.11.2015

# Contents

- Principle of FDTD
  - Derivation
- Basic properties
  - Stability
  - Dispersion
  - Boundary conditions
- Advantages and weaknesses
- Examples

# FDTD (= finite-difference time-domain)

## Principle

1. Start from Maxwell's equations
2. Replace all derivatives with finite-difference approximations
3. Done

# Maxwell's curl equations

$$\begin{aligned}\nabla \times \bar{E} &= -\mu \frac{\partial}{\partial t} \bar{H}, \\ \nabla \times \bar{H} &= \bar{J} + \sigma \bar{E} + \epsilon \frac{\partial}{\partial t} \bar{E},\end{aligned}\quad \left( \bar{H} = \frac{1}{\mu} \bar{B} \right)$$

$$\begin{cases} \mu \frac{\partial}{\partial t} H_x = \frac{\partial}{\partial z} E_y - \frac{\partial}{\partial y} E_z \\ \mu \frac{\partial}{\partial t} H_y = \frac{\partial}{\partial x} E_z - \frac{\partial}{\partial z} E_x \\ \mu \frac{\partial}{\partial t} H_z = \frac{\partial}{\partial y} E_x - \frac{\partial}{\partial x} E_y \end{cases}$$

$$\begin{cases} \epsilon \frac{\partial}{\partial t} E_x = \frac{\partial}{\partial y} H_z - \frac{\partial}{\partial z} H_y - \sigma E_x \\ \epsilon \frac{\partial}{\partial t} E_y = \frac{\partial}{\partial z} H_x - \frac{\partial}{\partial x} H_z - \sigma E_y \\ \epsilon \frac{\partial}{\partial t} E_z = \frac{\partial}{\partial x} H_y - \frac{\partial}{\partial y} H_x - \sigma E_z \end{cases}$$

# Central difference approximations

Notation

$$x = i \Delta x, y = j \Delta y, z = k \Delta z \quad t = n \Delta t$$

$$F(x, y, z, t) = F(i \Delta x, j \Delta y, k \Delta z, n \Delta t) \Rightarrow F^n(i, j, k)$$

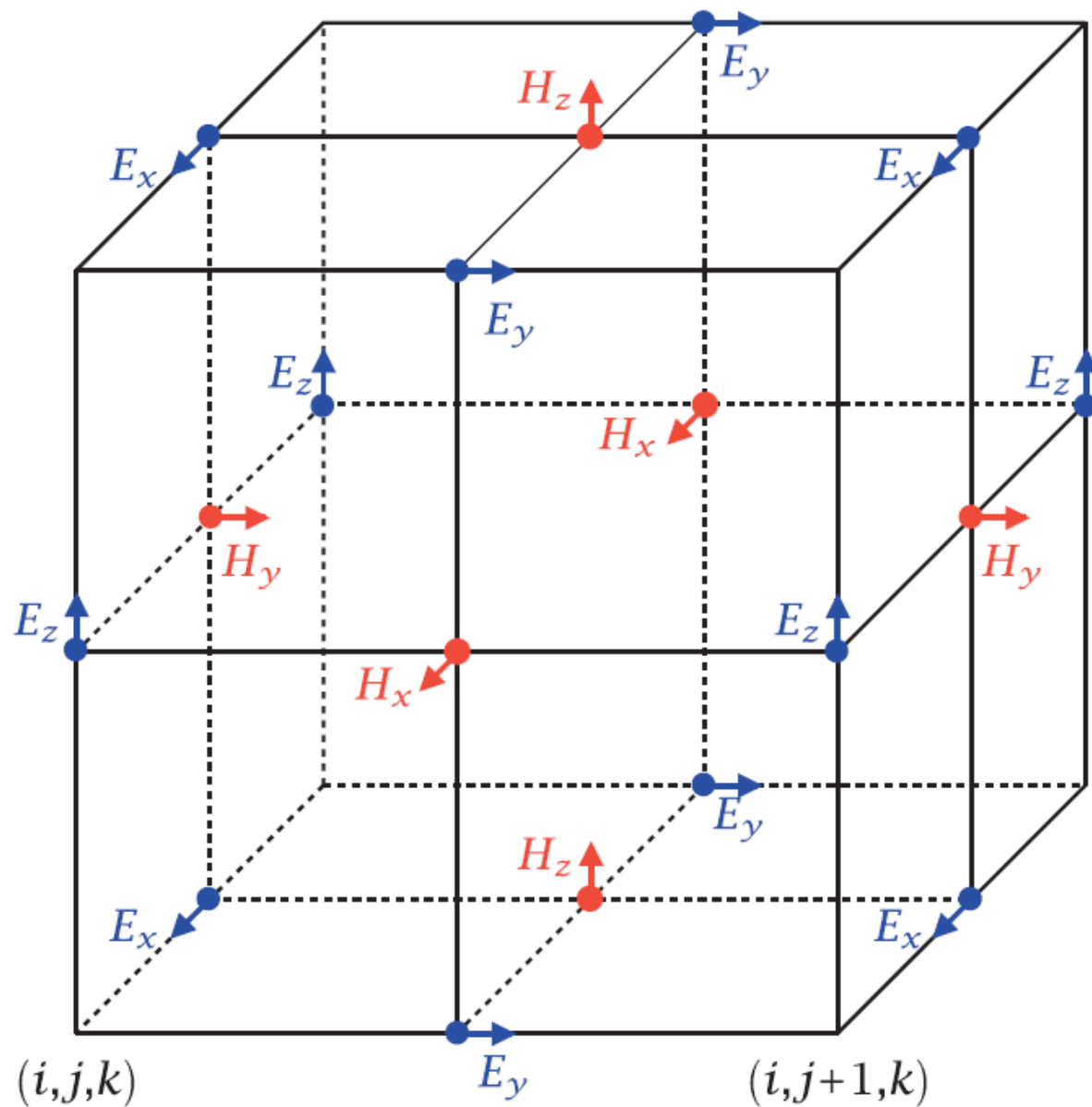
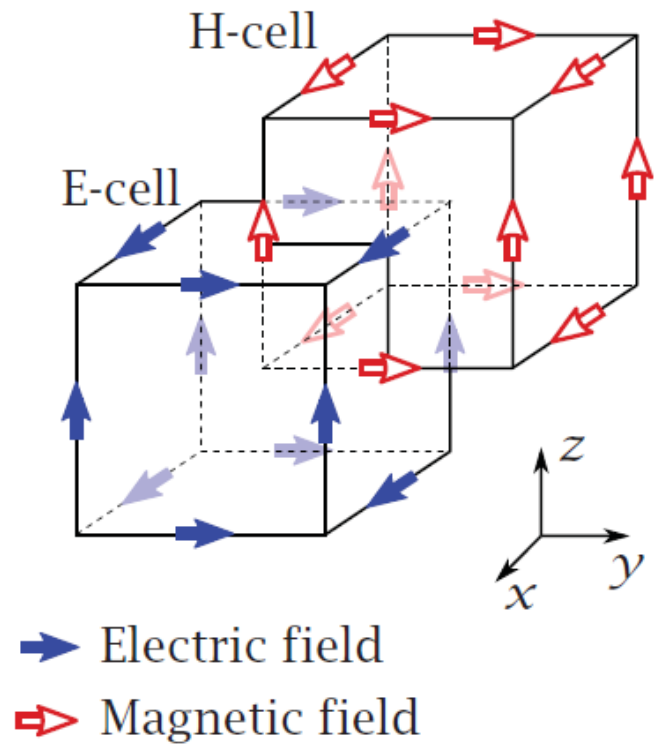
$$\frac{\partial F(x, y, z, t)}{\partial x} \approx \frac{F^n(i + \frac{1}{2}, j, k) - F^n(i - \frac{1}{2}, j, k)}{\Delta x} \quad O(\Delta x^2)$$

$$\frac{\partial F(x, y, z, t)}{\partial y} \approx \frac{F^n(i, j + \frac{1}{2}, k) - F^n(i, j - \frac{1}{2}, k)}{\Delta y} \quad O(\Delta y^2)$$

$$\frac{\partial F(x, y, z, t)}{\partial z} \approx \frac{F^n(i, j, k + \frac{1}{2}) - F^n(i, j, k - \frac{1}{2})}{\Delta z} \quad O(\Delta z^2)$$

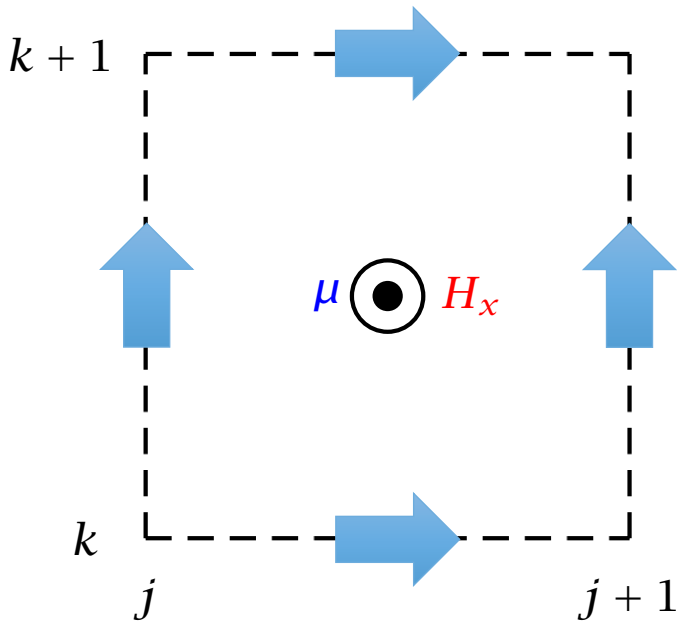
$$\frac{\partial F(x, y, z, t)}{\partial t} \approx \frac{F^{n+\frac{1}{2}}(i, j, k) - F^{n-\frac{1}{2}}(i, j, k)}{\Delta t} \quad O(\Delta t^2)$$

# Yee cell (Yee, 1966)



# Derivation of FDTD update equations

$$\mu \frac{\partial}{\partial t} H_x = \frac{\partial}{\partial z} E_y - \frac{\partial}{\partial y} E_z$$



$$\mu = \mu(i, j + \frac{1}{2}, k + \frac{1}{2})$$

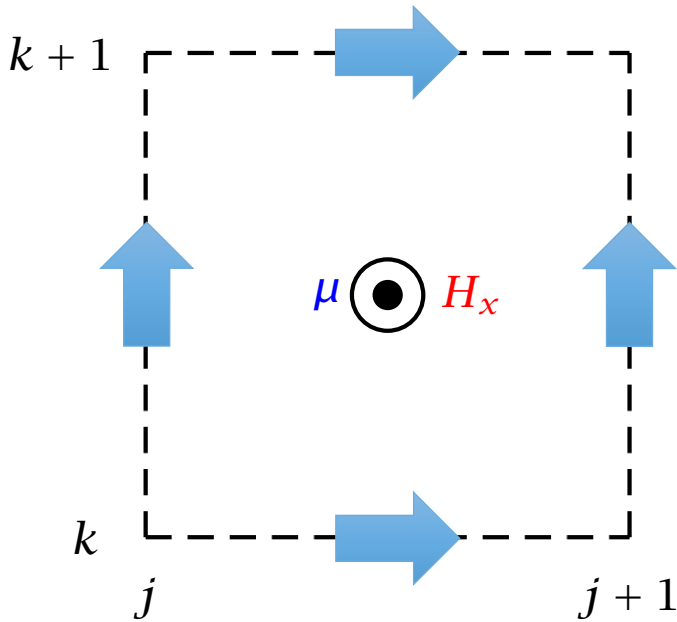
$$\frac{\partial}{\partial t} H_x^n(i, j + \frac{1}{2}, k + \frac{1}{2}) \approx \frac{H_x^{n+\frac{1}{2}}(i, j + \frac{1}{2}, k + \frac{1}{2}) - H_x^{n-\frac{1}{2}}(i, j + \frac{1}{2}, k + \frac{1}{2})}{\Delta t}$$

$$\frac{\partial}{\partial z} E_y^n(i, j + \frac{1}{2}, k + \frac{1}{2}) \approx \frac{E_y^n(i, j + \frac{1}{2}, k + 1) - E_y^n(i, j + \frac{1}{2}, k)}{\Delta z}$$

$$\frac{\partial}{\partial y} E_z^n(i, j + \frac{1}{2}, k + \frac{1}{2}) \approx \frac{E_z^n(i, j + 1, k + \frac{1}{2}) - E_z^n(i, j, k + \frac{1}{2})}{\Delta y}$$

# Derivation of FDTD update equations

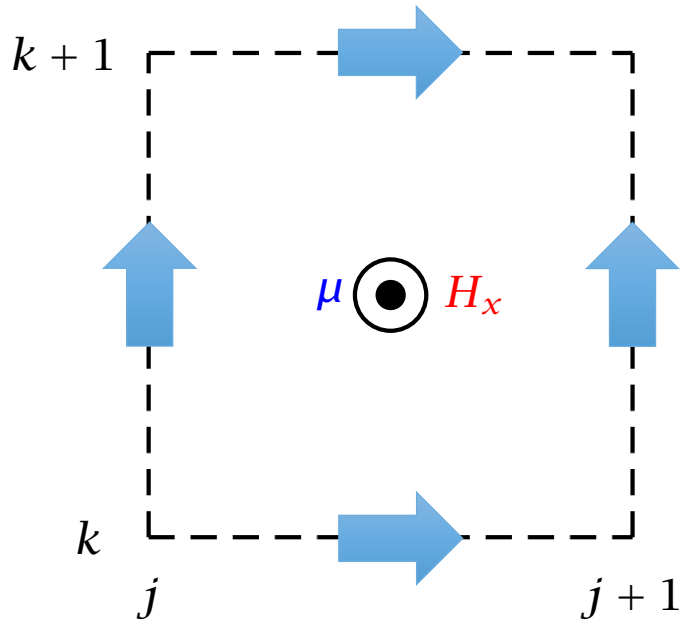
$$\mu \frac{\partial}{\partial t} H_x = \frac{\partial}{\partial z} E_y - \frac{\partial}{\partial y} E_z$$



$$\begin{aligned} & \mu(i, j + \frac{1}{2}, k + \frac{1}{2}) \frac{H_x^{n+\frac{1}{2}}(i, j + \frac{1}{2}, k + \frac{1}{2}) - H_x^{n-\frac{1}{2}}(i, j + \frac{1}{2}, k + \frac{1}{2})}{\Delta t} \\ &= \frac{E_y^n(i, j + \frac{1}{2}, k + 1) - E_y^n(i, j + \frac{1}{2}, k)}{\Delta z} - \frac{E_z^n(i, j + 1, k + \frac{1}{2}) - E_z^n(i, j, k + \frac{1}{2})}{\Delta y} \end{aligned}$$



# FDTD update equations



$$H_x^{n+\frac{1}{2}}(i, j + \frac{1}{2}, k + \frac{1}{2}) = H_x^{n-\frac{1}{2}}(i, j + \frac{1}{2}, k + \frac{1}{2})$$

$$+ \frac{\Delta t}{\mu(i, j + \frac{1}{2}, k + \frac{1}{2})\Delta z} [E_y^n(i, j + \frac{1}{2}, k + 1) - E_y^n(i, j + \frac{1}{2}, k)]$$

$$+ \frac{\Delta t}{\mu(i, j + \frac{1}{2}, k + \frac{1}{2})\Delta y} [E_z^n(i, j, k + \frac{1}{2}) - E_z^n(i, j + 1, k + \frac{1}{2})]$$

$$H_y^{n+\frac{1}{2}}(i + \frac{1}{2}, j, k + \frac{1}{2}) = H_y^{n-\frac{1}{2}}(i + \frac{1}{2}, j, k + \frac{1}{2})$$

$$+ \frac{\Delta t}{\mu(i + \frac{1}{2}, j, k + \frac{1}{2})\Delta x} [E_z^n(i + 1, j, k + \frac{1}{2}) - E_z^n(i, j, k + \frac{1}{2})]$$

$$+ \frac{\Delta t}{\mu(i + \frac{1}{2}, j, k + \frac{1}{2})\Delta z} [E_x^n(i + \frac{1}{2}, j, k) - E_x^n(i + \frac{1}{2}, j, k + 1)]$$

# FDTD update equations

$$H_z^{n+\frac{1}{2}}(i+\frac{1}{2}, j+\frac{1}{2}, k) = H_z^{n-\frac{1}{2}}(i+\frac{1}{2}, j+\frac{1}{2}, k)$$

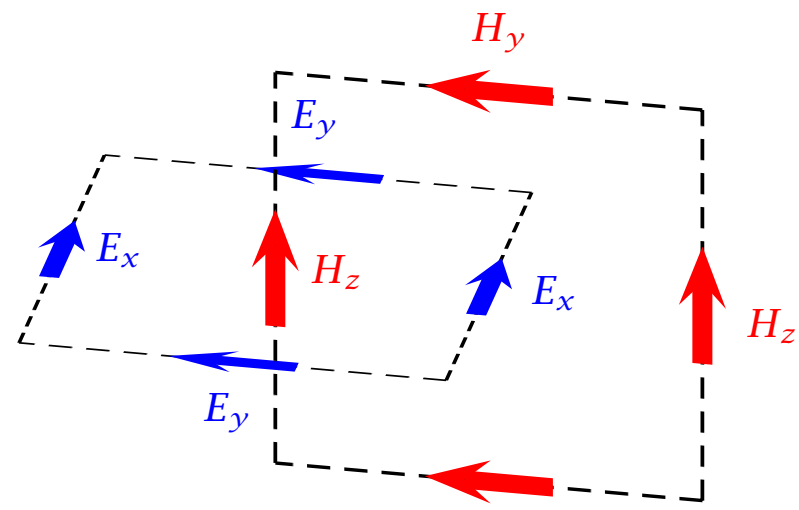
$$+ \frac{\Delta t}{\mu(i+\frac{1}{2}, j+\frac{1}{2}, k)\Delta y} [E_x^n(i+\frac{1}{2}, j+1, k) - E_x^n(i+\frac{1}{2}, j, k)]$$

$$+ \frac{\Delta t}{\mu(i+\frac{1}{2}, j+\frac{1}{2}, k)\Delta x} [E_y^n(i, j+\frac{1}{2}, k) - E_y^n(i+1, j+\frac{1}{2}, k)]$$

$$E_x^{n+1}(i+\frac{1}{2}, j, k) = \frac{2\epsilon(i+\frac{1}{2}, j, k) - \sigma(i+\frac{1}{2}, j, k)\Delta t}{2\epsilon(i+\frac{1}{2}, j, k) + \sigma(i+\frac{1}{2}, j, k)\Delta t} E_x^n(i+\frac{1}{2}, j, k)$$

$$+ \frac{2\Delta t}{[2\epsilon(i+\frac{1}{2}, j, k) + \sigma(i+\frac{1}{2}, j, k)\Delta t]\Delta y} \left[ H_z^{n+\frac{1}{2}}(i+\frac{1}{2}, j+\frac{1}{2}, k) - H_z^{n+\frac{1}{2}}(i+\frac{1}{2}, j-\frac{1}{2}, k) \right]$$

$$+ \frac{2\Delta t}{[2\epsilon(i+\frac{1}{2}, j, k) + \sigma(i+\frac{1}{2}, j, k)\Delta t]\Delta z} \left[ H_y^{n+\frac{1}{2}}(i+\frac{1}{2}, j, k-\frac{1}{2}) - H_y^{n+\frac{1}{2}}(i+\frac{1}{2}, j, k+\frac{1}{2}) \right]$$



# FDTD update equations

$$E_y^{n+1}(i, j + \frac{1}{2}, k) = \frac{2\epsilon(i, j + \frac{1}{2}, k) - \sigma(i, j + \frac{1}{2}, k)\Delta t}{2\epsilon(i, j + \frac{1}{2}, k) + \sigma(i, j + \frac{1}{2}, k)\Delta t} E_y^n(i, j + \frac{1}{2}, k)$$

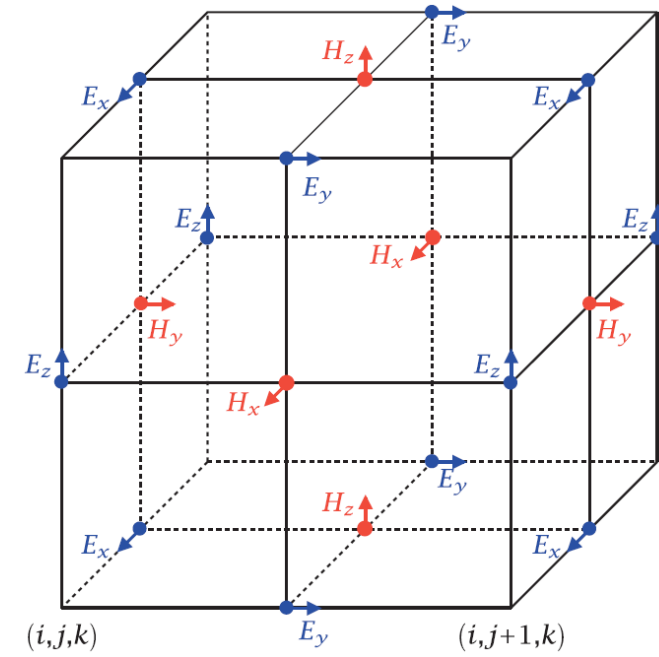
$$+ \frac{2\Delta t}{[2\epsilon(i, j + \frac{1}{2}, k) + \sigma(i, j + \frac{1}{2}, k)\Delta t]\Delta z} \left[ H_x^{n+\frac{1}{2}}(i, j + \frac{1}{2}, k + \frac{1}{2}) - H_x^{n+\frac{1}{2}}(i, j + \frac{1}{2}, k - \frac{1}{2}) \right]$$

$$+ \frac{2\Delta t}{[2\epsilon(i, j + \frac{1}{2}, k) + \sigma(i, j + \frac{1}{2}, k)\Delta t]\Delta x} \left[ H_z^{n+\frac{1}{2}}(i - \frac{1}{2}, j + \frac{1}{2}, k) - H_z^{n+\frac{1}{2}}(i + \frac{1}{2}, j + \frac{1}{2}, k) \right]$$

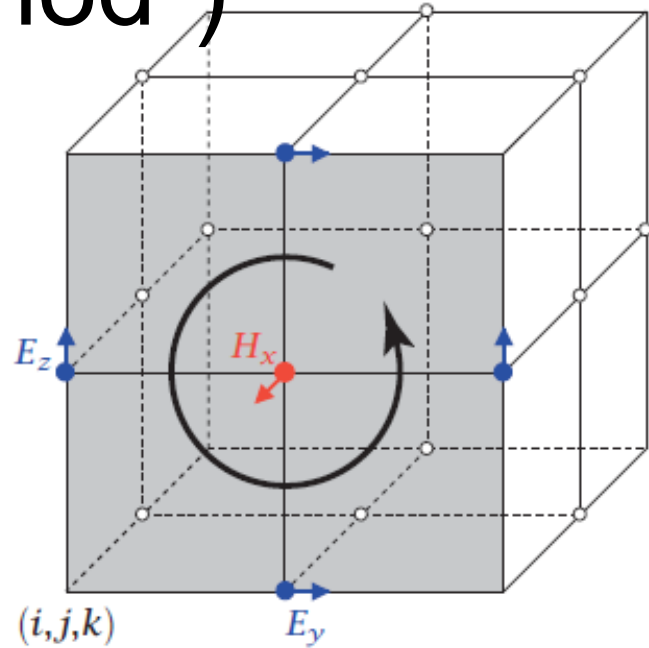
$$E_z^{n+1}(i, j, k + \frac{1}{2}) = \frac{2\epsilon(i, j, k + \frac{1}{2}) - \sigma(i, j, k + \frac{1}{2})\Delta t}{2\epsilon(i, j, k + \frac{1}{2}) + \sigma(i, j, k + \frac{1}{2})\Delta t} E_z^n(i, j, k + \frac{1}{2})$$

$$+ \frac{2\Delta t}{[2\epsilon(i, j, k + \frac{1}{2}) + \sigma(i, j, k + \frac{1}{2})\Delta t]\Delta x} \left[ H_y^{n+\frac{1}{2}}(i + \frac{1}{2}, j, k + \frac{1}{2}) - H_y^{n+\frac{1}{2}}(i - \frac{1}{2}, j, k + \frac{1}{2}) \right]$$

$$+ \frac{2\Delta t}{[2\epsilon(i, j, k + \frac{1}{2}) + \sigma(i, j, k + \frac{1}{2})\Delta t]\Delta y} \left[ H_x^{n+\frac{1}{2}}(i, j - \frac{1}{2}, k + \frac{1}{2}) - H_x^{n+\frac{1}{2}}(i, j + \frac{1}{2}, k + \frac{1}{2}) \right]$$



# Derivation of FDTD update equations from integral form of Maxwell's equations ("FIT method")



$$\nabla \times \vec{E} = -\mu \frac{\partial}{\partial t} \vec{H}$$

$\Downarrow$

$$\int_S \nabla \times \vec{E} \cdot \vec{dS} = - \int_S \mu \frac{\partial}{\partial t} \vec{H} \cdot \vec{dS}$$

$\Downarrow$

$$\int_{\partial S} \vec{E} \cdot \vec{dc} = - \int_S \mu \frac{\partial}{\partial t} \vec{H} \cdot \vec{dS}$$

Use the mid-ordinate numerical integration method

$$\int_{\partial S} \vec{E} \cdot \vec{dc} \approx E_y^n(i, j + \frac{1}{2}, k) \Delta y + E_z^n(i, j + 1, k + \frac{1}{2}) \Delta z - E_y^n(i, j + \frac{1}{2}, k + 1) \Delta y - E_z^n(i, j, k + \frac{1}{2}) \Delta z$$

$$- \int_S \mu \frac{\partial}{\partial t} \vec{H} \cdot \vec{dS} \approx -\mu(i, j + \frac{1}{2}, k + \frac{1}{2}) \frac{\left[ H_x^{n+\frac{1}{2}}(i, j + \frac{1}{2}, k + \frac{1}{2}) - H_x^{n-\frac{1}{2}}(i, j + \frac{1}{2}, k + \frac{1}{2}) \right]}{\Delta t} \Delta y \Delta z$$

# Example: 1D FDTD

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

$$\epsilon \frac{\partial}{\partial t} E_z = \frac{\partial}{\partial x} H_y$$

$$H_y^{n+\frac{1}{2}}(i + \frac{1}{2}) = H_y^{n-\frac{1}{2}}(i + \frac{1}{2}) + \frac{\Delta t}{\mu(i + \frac{1}{2})\Delta x} (E_z^n(i + 1) - E_z^n(i))$$

$$E_z^{n+1}(i) = E_z(i)^n + \frac{\Delta t}{\epsilon(i)\Delta x} \left( H_y^{n+\frac{1}{2}}(i + \frac{1}{2}) - H_y^{n+\frac{1}{2}}(i - \frac{1}{2}) \right)$$

# Implementation in MATLAB

$$H_y^{n+\frac{1}{2}}(i + \frac{1}{2}) = H_y^{n-\frac{1}{2}}(i + \frac{1}{2}) + \frac{\Delta t}{\mu(i + \frac{1}{2})\Delta x} (E_z^n(i + 1) - E_z^n(i))$$

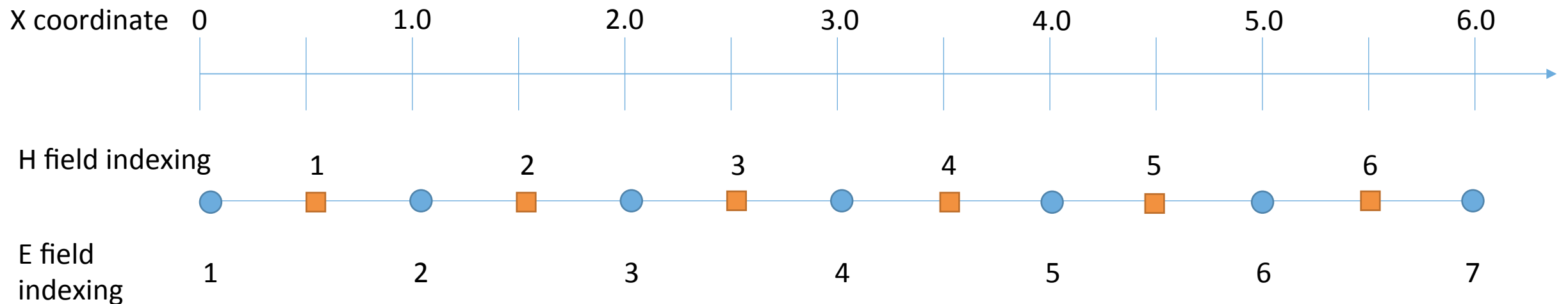
$$E_z^{n+1}(i) = E_z(i)^n + \frac{\Delta t}{\epsilon(i)\Delta x} \left( H_y^{n+\frac{1}{2}}(i + \frac{1}{2}) - H_y^{n+\frac{1}{2}}(i - \frac{1}{2}) \right)$$

% Magnetic field update equation

```
Hy(1:K-1) = Hy(1:K-1) + Db(1:K-1) .* ( Ez(2:K) - Ez(1:K-1) );
```

% Electric field update equation

```
Ez(2:K-1) = Ez(2:K-1) + Cb(2:K-1) .* ( Hy(2:K-1) - Hy(1:K-2) );
```



# Contents

- Principle of FDTD
  - Derivation
- Basic properties
  - Stability
  - Dispersion
  - Boundary conditions
- Advantages and weaknesses
- Examples

# Stability

- Courant-Friedrichs-Lewy (CFL) condition:

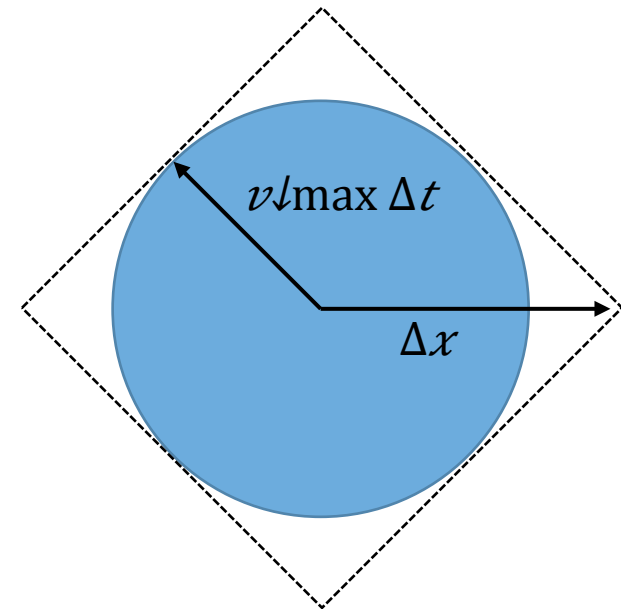
$$\Delta t \leq \frac{1}{v_{\max}} \frac{1}{\sqrt{\frac{1}{\Delta x^2} + \frac{1}{\Delta y^2} + \frac{1}{\Delta z^2}}}$$

$$v_{\max} = \left\{ \frac{1}{\sqrt{\epsilon \mu}} \right\}_{\max}$$

In 3D:

$$\Delta x = \Delta y = \Delta z = \Delta$$
$$\Delta t \leq \frac{\Delta}{\sqrt{3} v_{\max}}$$

Numerical domain of dependence must include analytical domain of dependence





# Numerical dispersion

Accurate formula

$$\frac{\omega^2}{c^2} = k_x^2 + k_y^2 + k_z^2$$

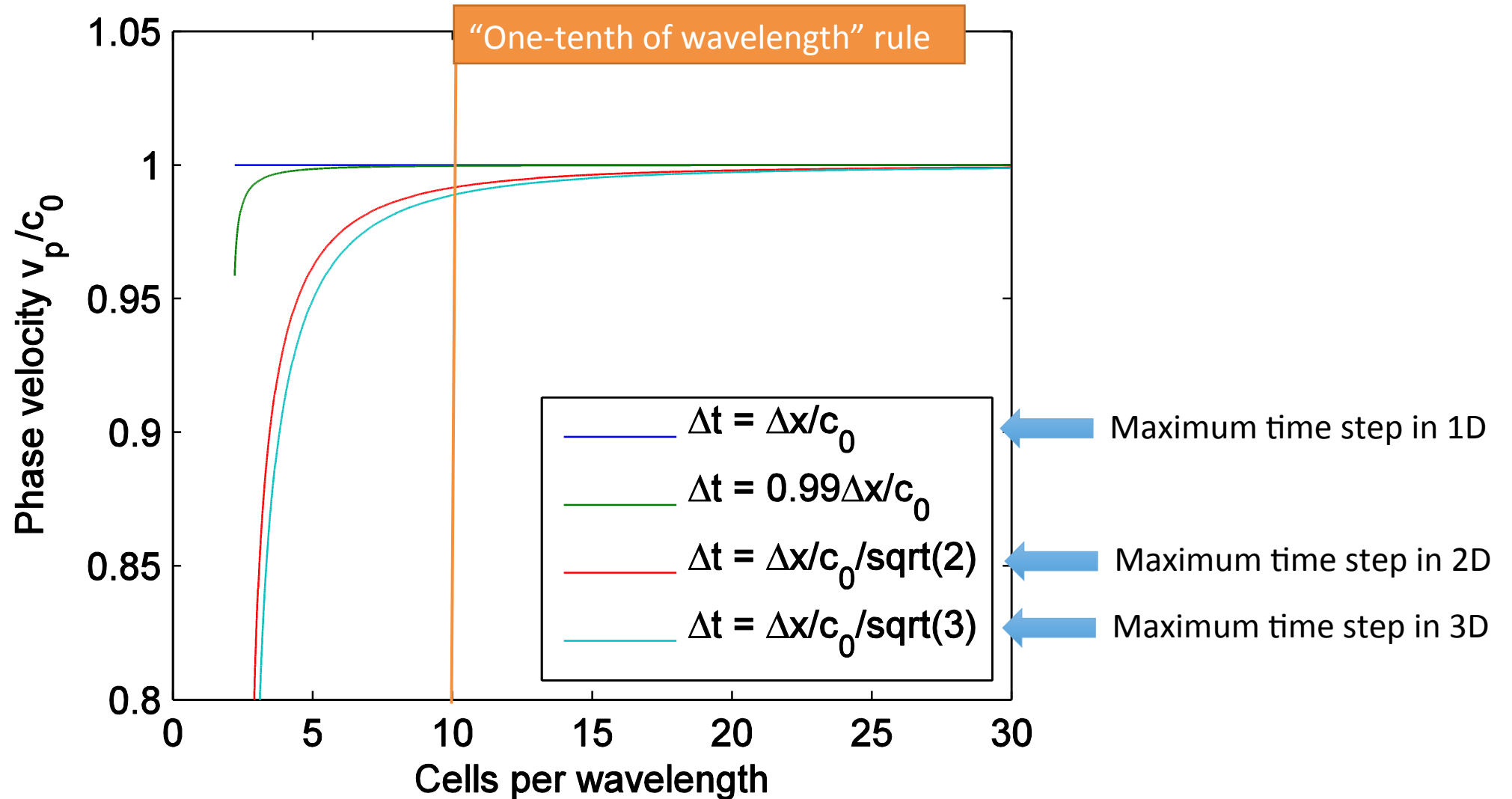
FDTD

$$\frac{\sin^2(\omega\Delta t/2)}{(c\Delta t)^2} = \frac{\sin^2(\tilde{k}_x\Delta x/2)}{(\Delta x)^2} + \frac{\sin^2(\tilde{k}_y\Delta y/2)}{(\Delta y)^2} + \frac{\sin^2(\tilde{k}_z\Delta z/2)}{(\Delta z)^2}$$

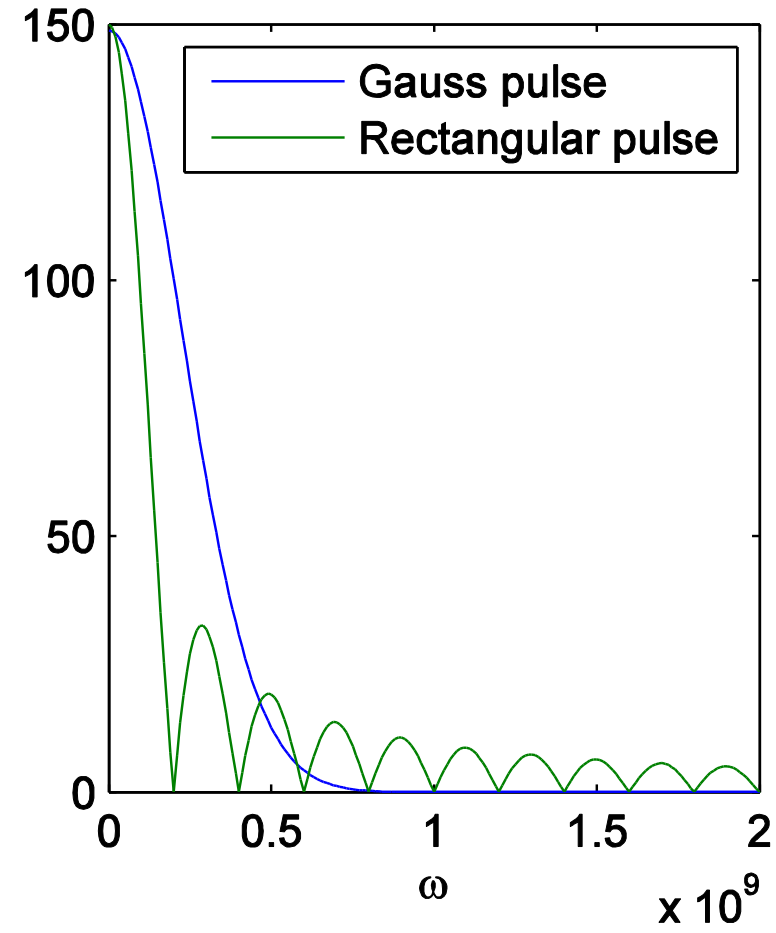
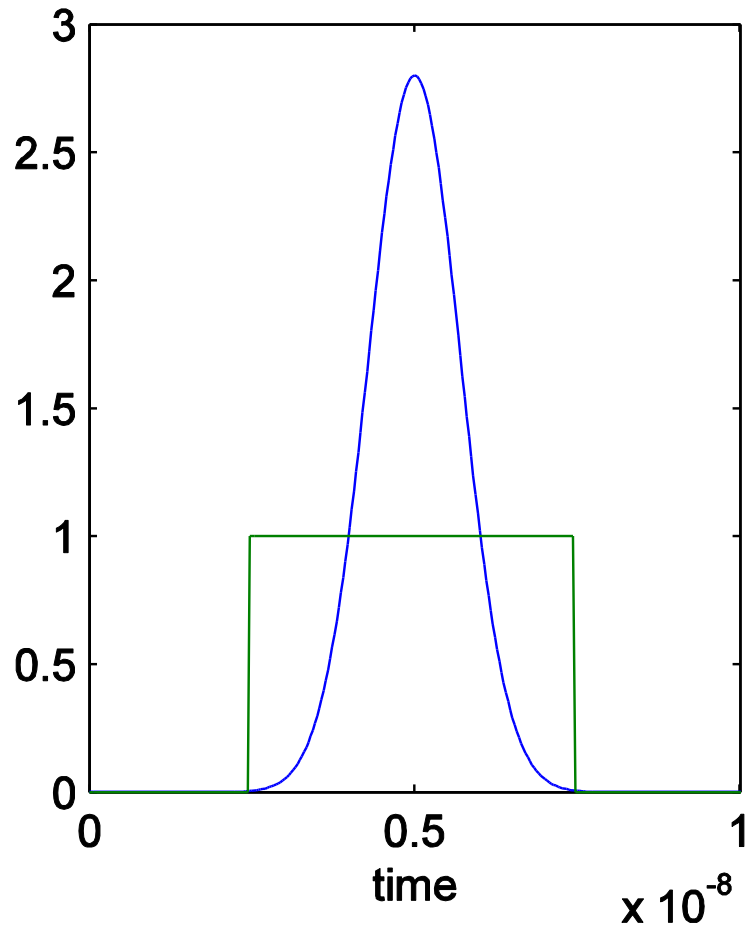
Numerical phase velocity

$$v_p = \frac{\omega}{\tilde{k}}$$

# Numerical dispersion

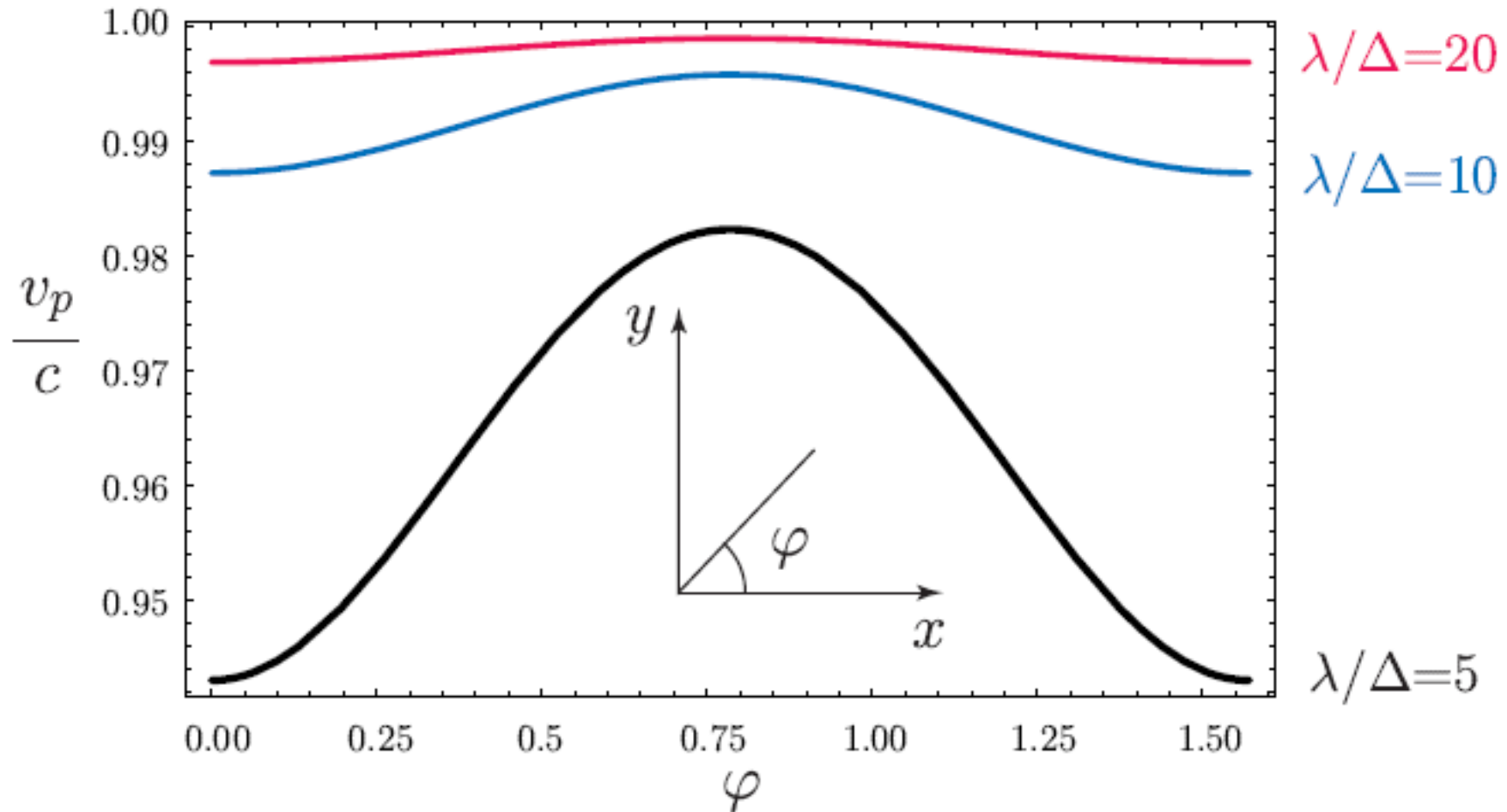


# Spectra



# Anisotropic dispersion in 2D and 3D

$$\Delta x = \Delta y = \Delta \quad c\Delta t = \frac{\Delta}{2} \leq \frac{\Delta}{\sqrt{3}}$$

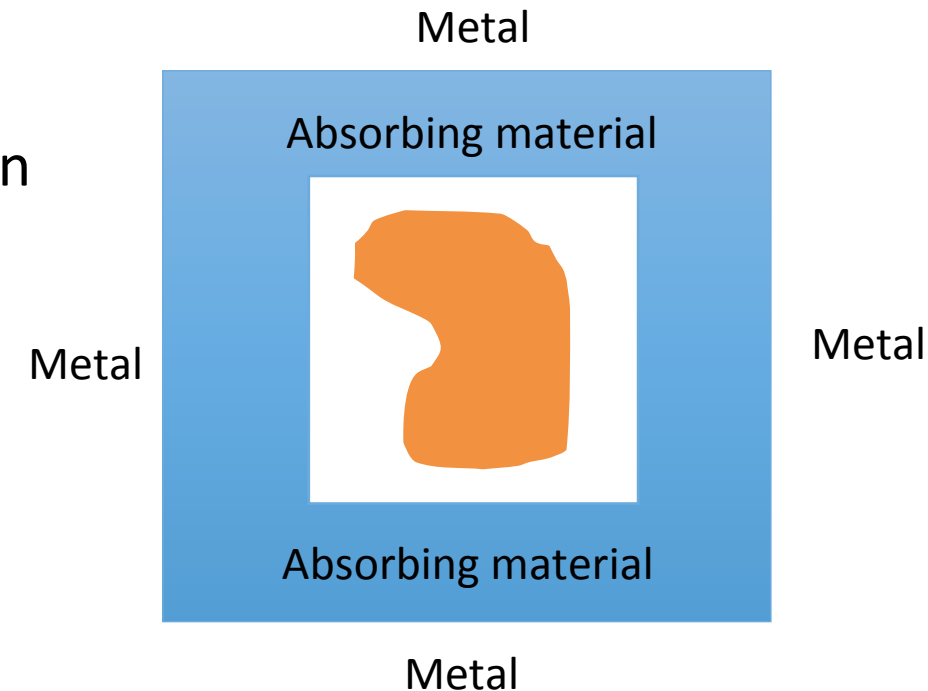
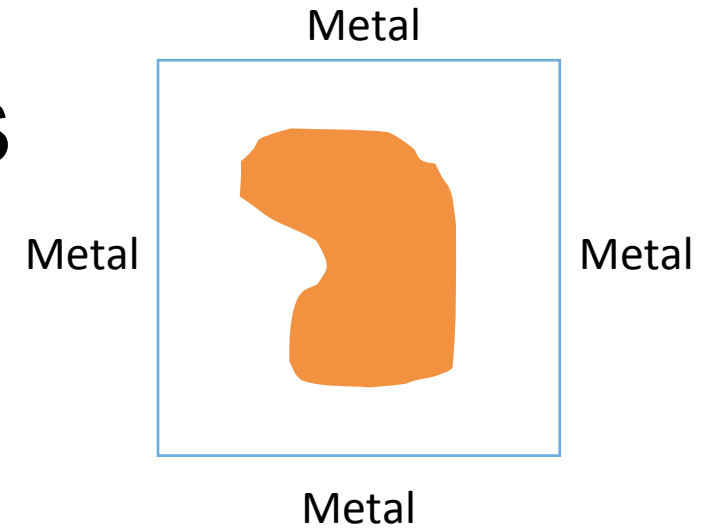


# Example: 2D FDTD

- 2D FDTD and anisotropic dispersion
  - example3a.m
- Be careful with dielectric materials
  - wavelength is shorter => finer cell size is needed
  - example3b.m

# Absorbing boundary conditions

- In the basic form, FDTD can only model boxes, with ideally conducting walls
  - How to terminate the computation domain to model free space?
- FDTD is very good at modelling different materials:
  - Berenger 1994: Make the walls of the box from an unphysical material that absorbs anything!
- ABCs are essential for any FDTD code
  - Example4.m, example5a, 5b, 5c



# Contents

- Principle of FDTD
  - Derivation
- Basic properties
  - Stability
  - Dispersion
  - Boundary conditions
- Advantages and weaknesses
- Applications

# Advantages of FDTD (1)

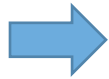
- Simple equations
- Can be parallelized easily
- Scales linearly with number of unknowns
- No need to solve equation systems

=> good for very large problems



# Microprocessor Transistor Counts 1971-2011 & Moore's Law

~ FDTD number of cells

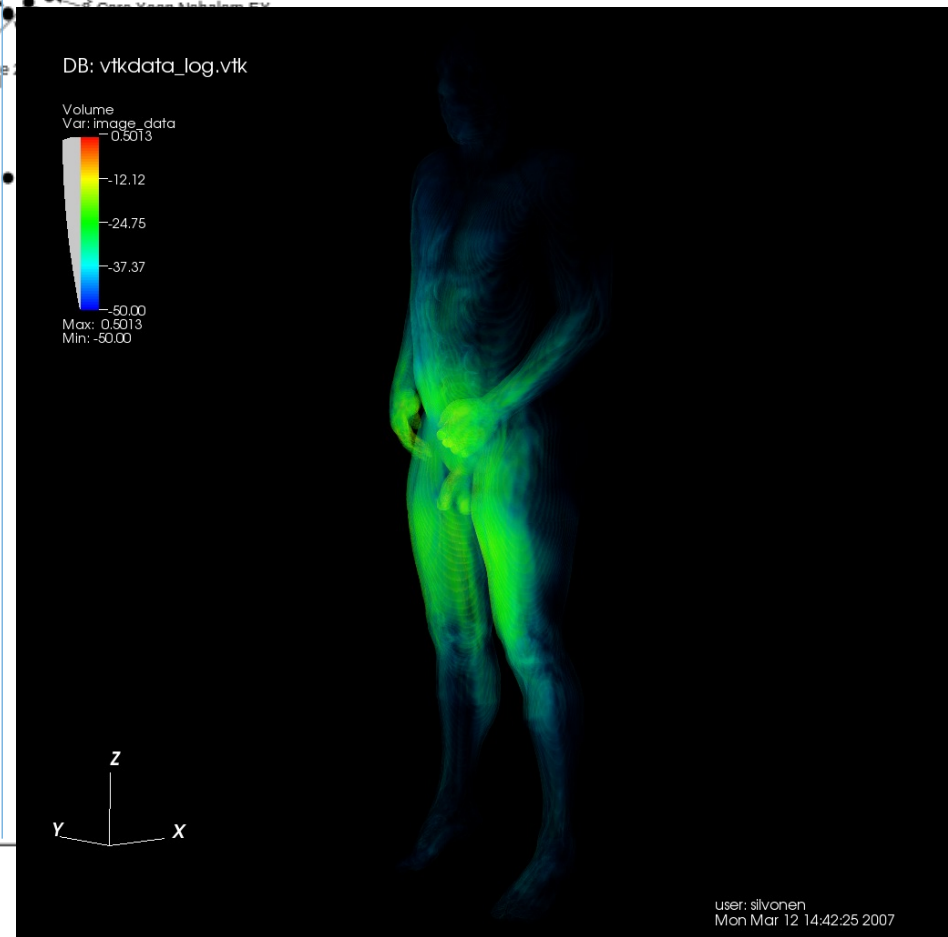
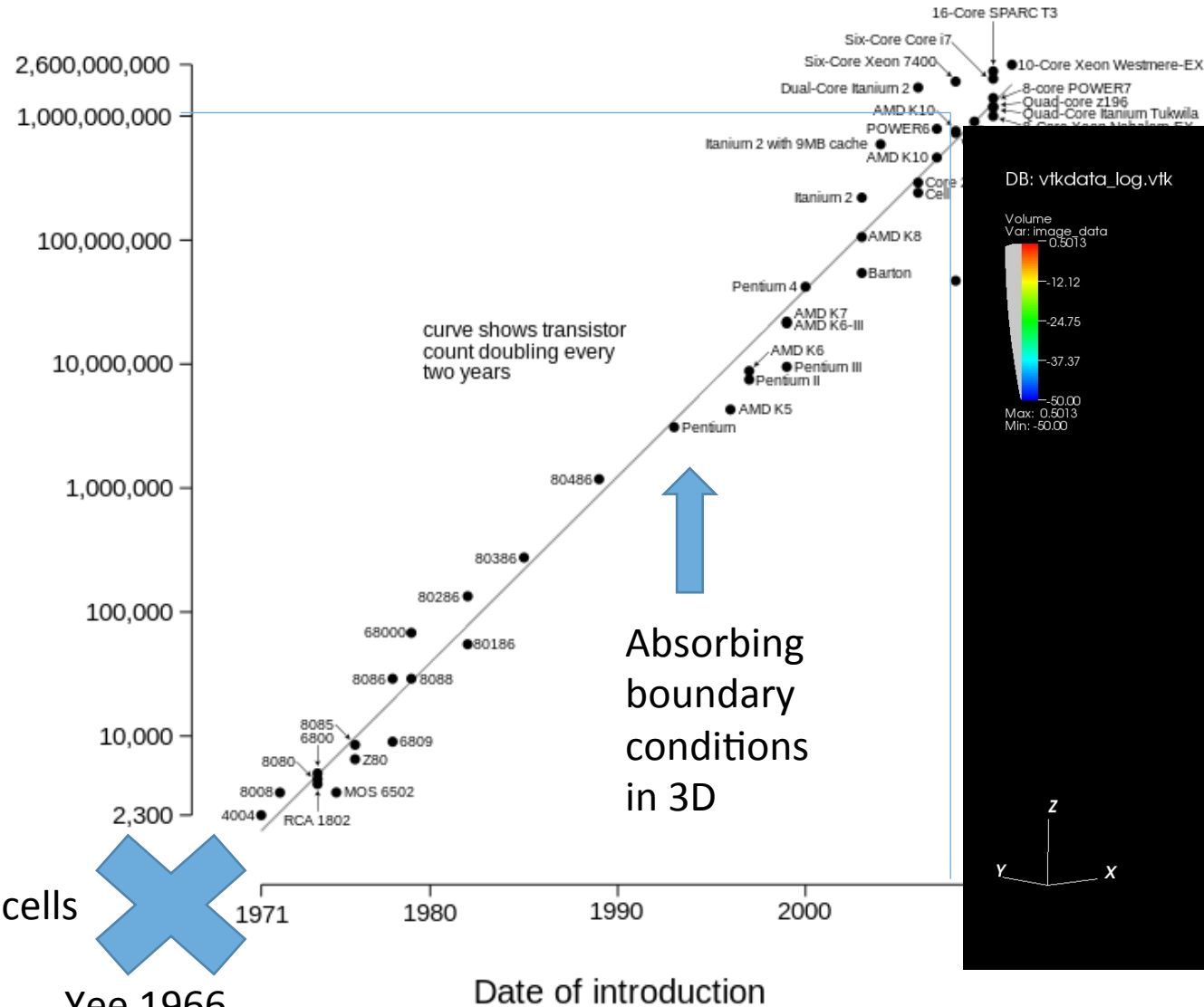


Transistor count

800 cells



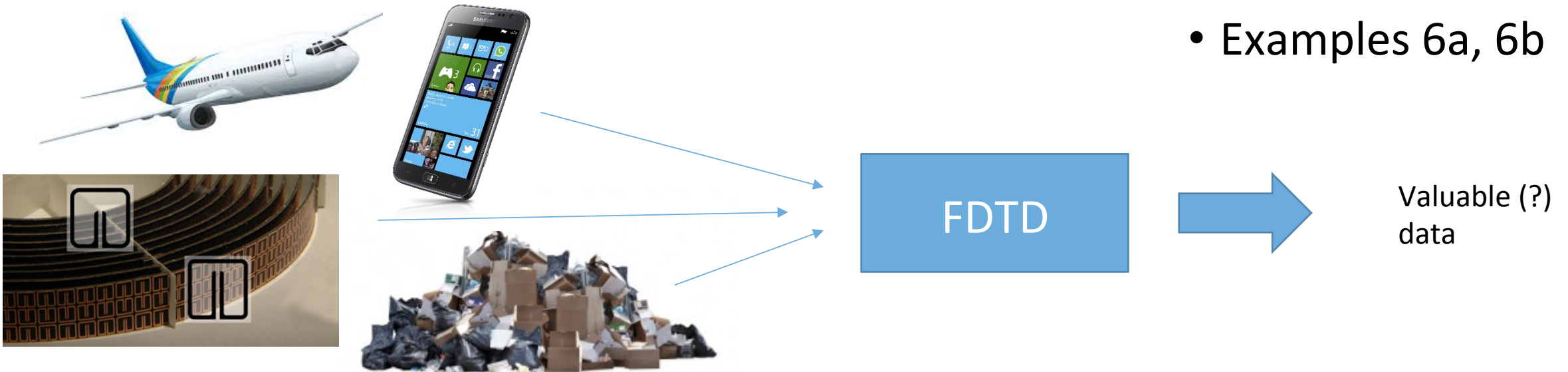
Yee 1966



# Advantages of FDTD (2)

- Any kind of time-domain sources
- Geometry and boundary conditions are taken into account automatically. Any shape can be modeled easily
- Different media can be modelled naturally: non-linearity, inhomogeneity, anisotropy, complex geometry (metamaterials)

- Examples 6a, 6b



# Weaknesses of FDTD (1)

- Not good for slow phenomena (huge number of time steps needed)

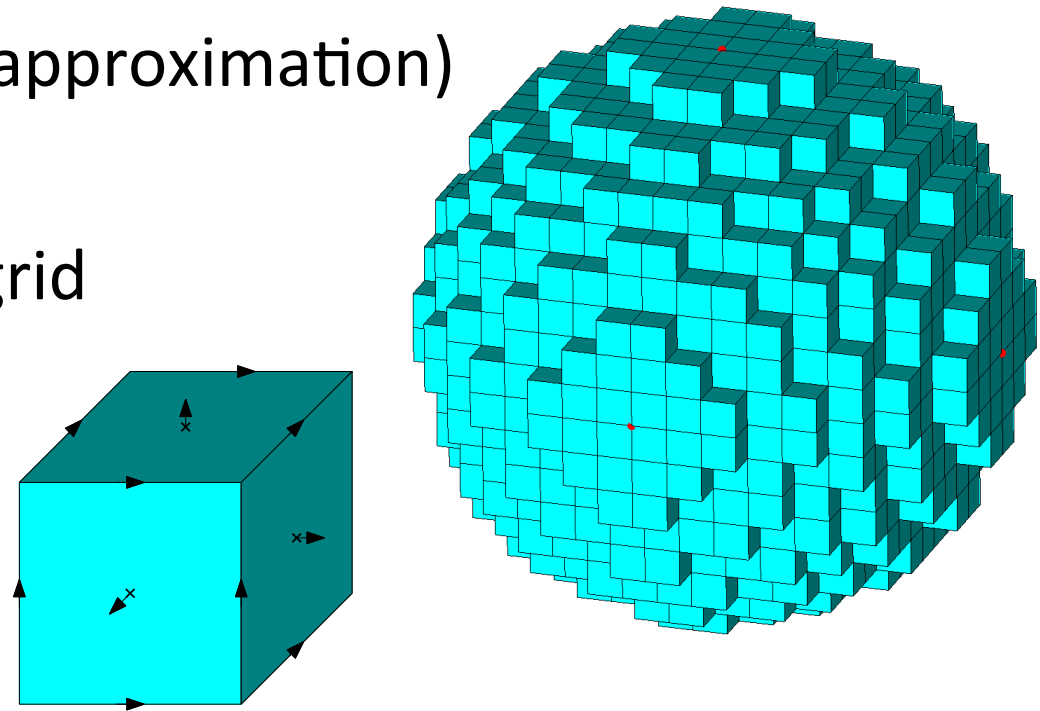
Example: 1 mm grid resolution  $\rightarrow$  time step =  $1.9\text{e-}12$  s

Phenomenon lasting 1 ms  $\rightarrow$  Number of time steps =  $5\text{e}8$

- Curved shapes are problematic (staircase approximation)

- High permittivity medium requires a fine grid

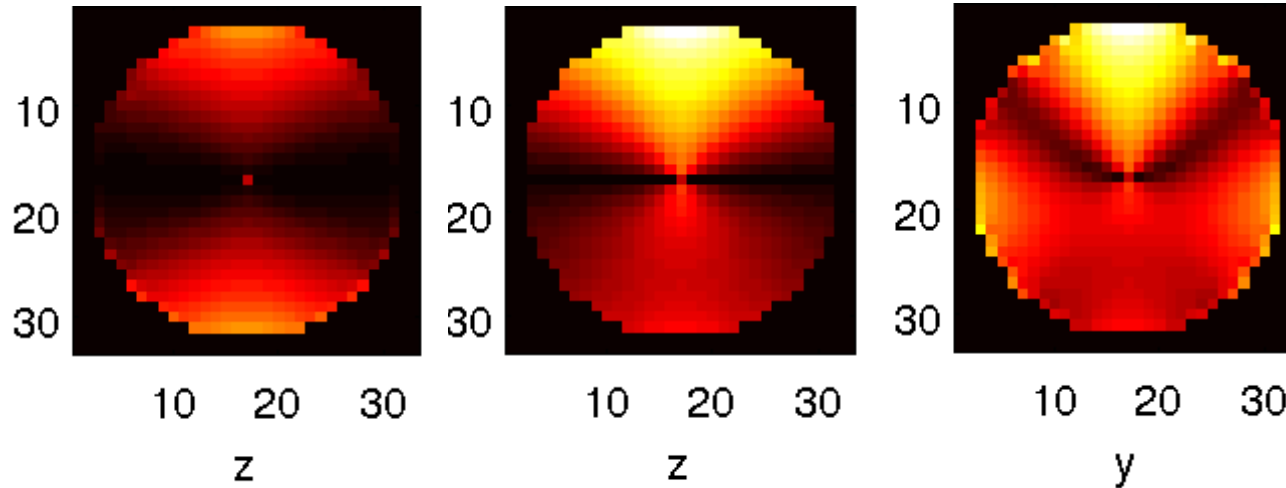
- One-tenth of wavelength rule



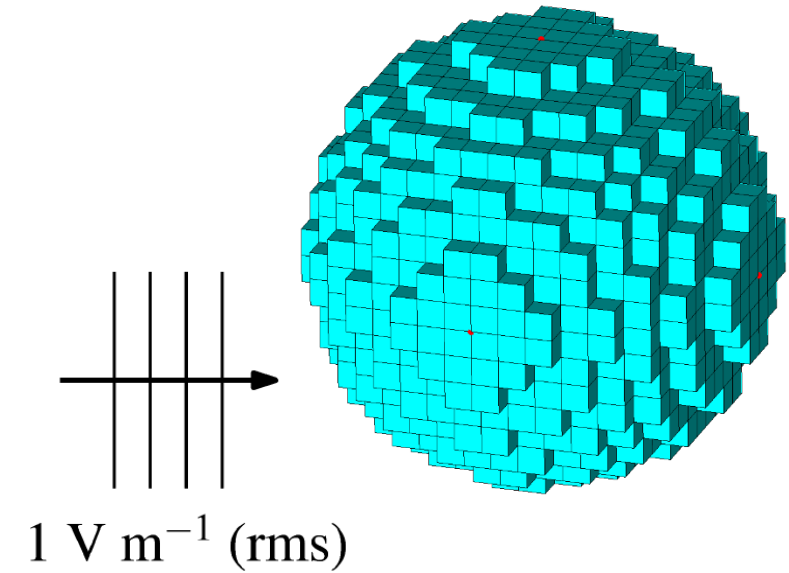
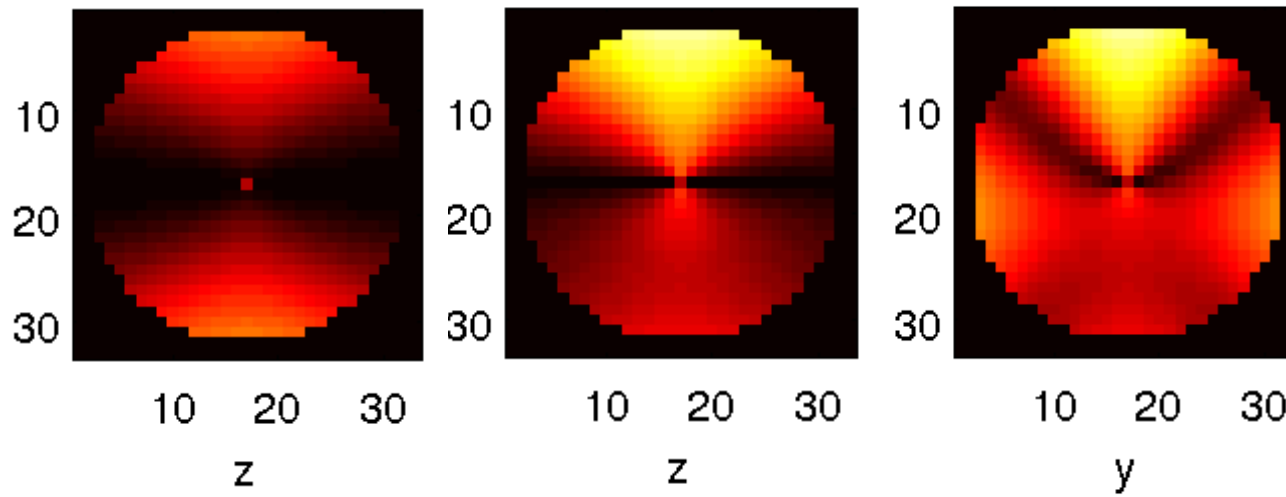
# Comparison with analytic solution

Radial component of the electric field  
in a sphere

FDTD

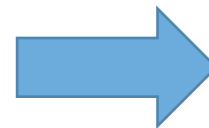
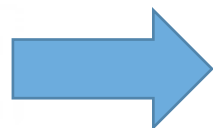
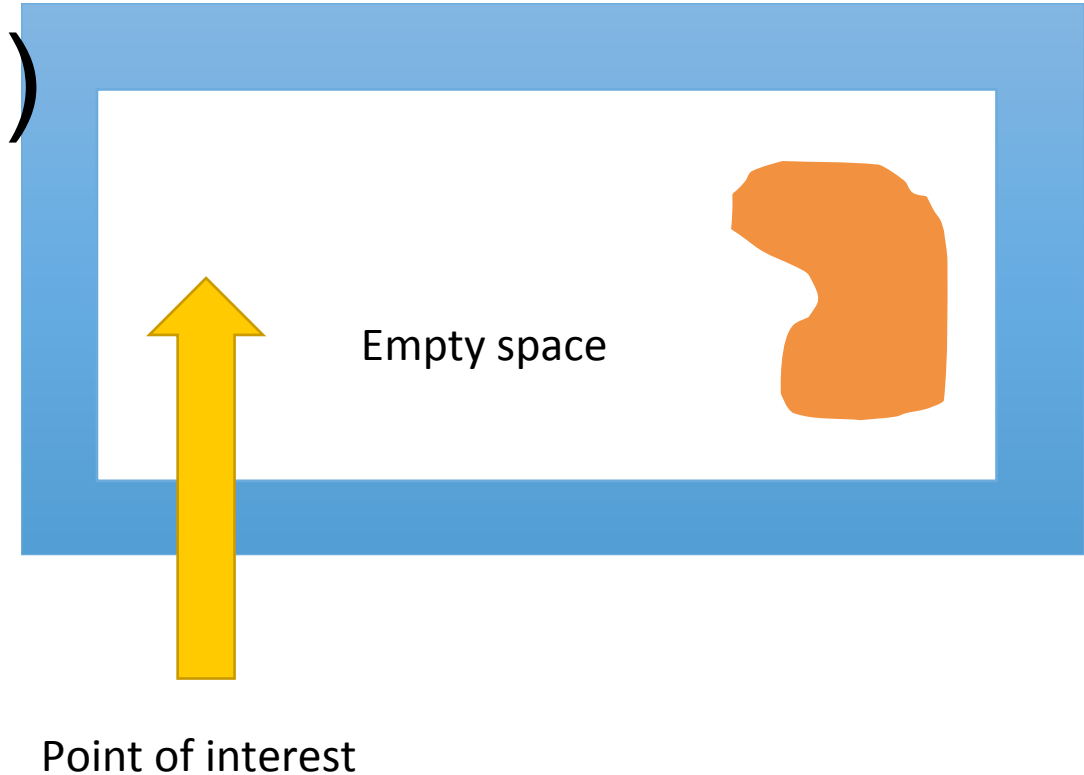


Analytic  
solution



# Weaknesses of FDTD (2)

- Computation domain must be finite
  - Absorbing boundary conditions
- Entire computational domain needs to be gridded (also empty space)
- Results depend on the choice of coordinate axes
- Error control



# Error in FDTD

- Truncation error from difference approximation (time and space)
  - Dispersion error, numerical anisotropy
  - Unphysical Poynting theorem, conservation of energy
  - ~~Floating point (round-off) error~~
  - Staircase approximation error
  - Absorbing boundary condition error
  - Modelling dielectric/lossy materials
- etc.

# Contents

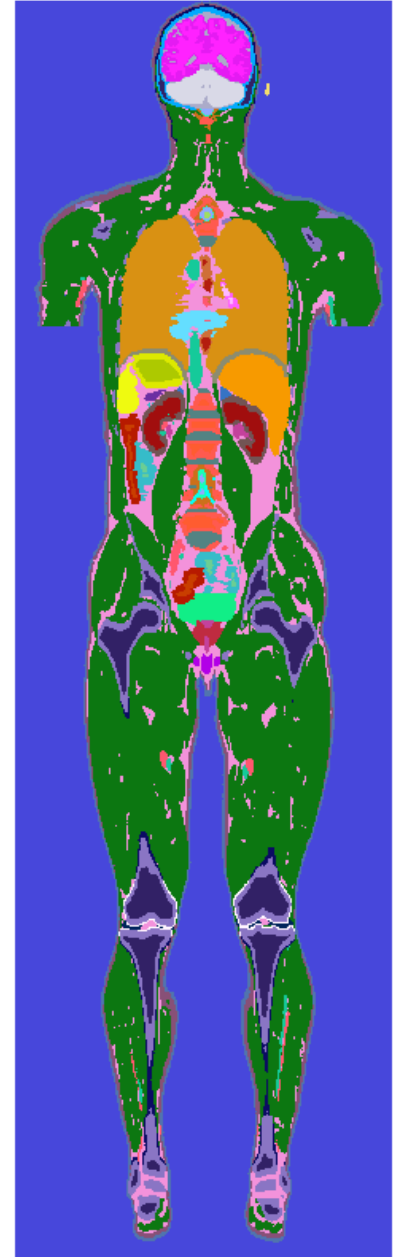
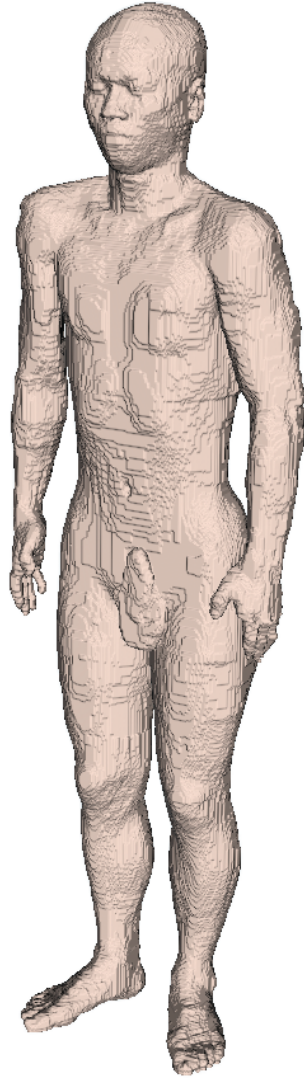
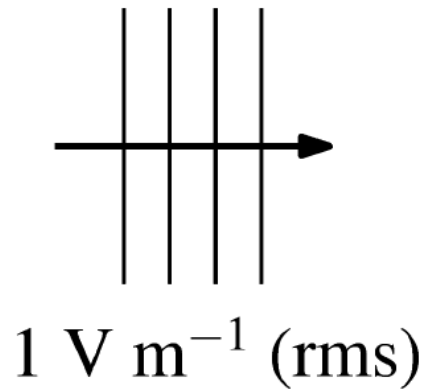
- Principle of FDTD
  - Derivation
- Basic properties
  - Stability
  - Dispersion
  - Boundary conditions
- Advantages and weaknesses
- Applications

# Applications of FDTD

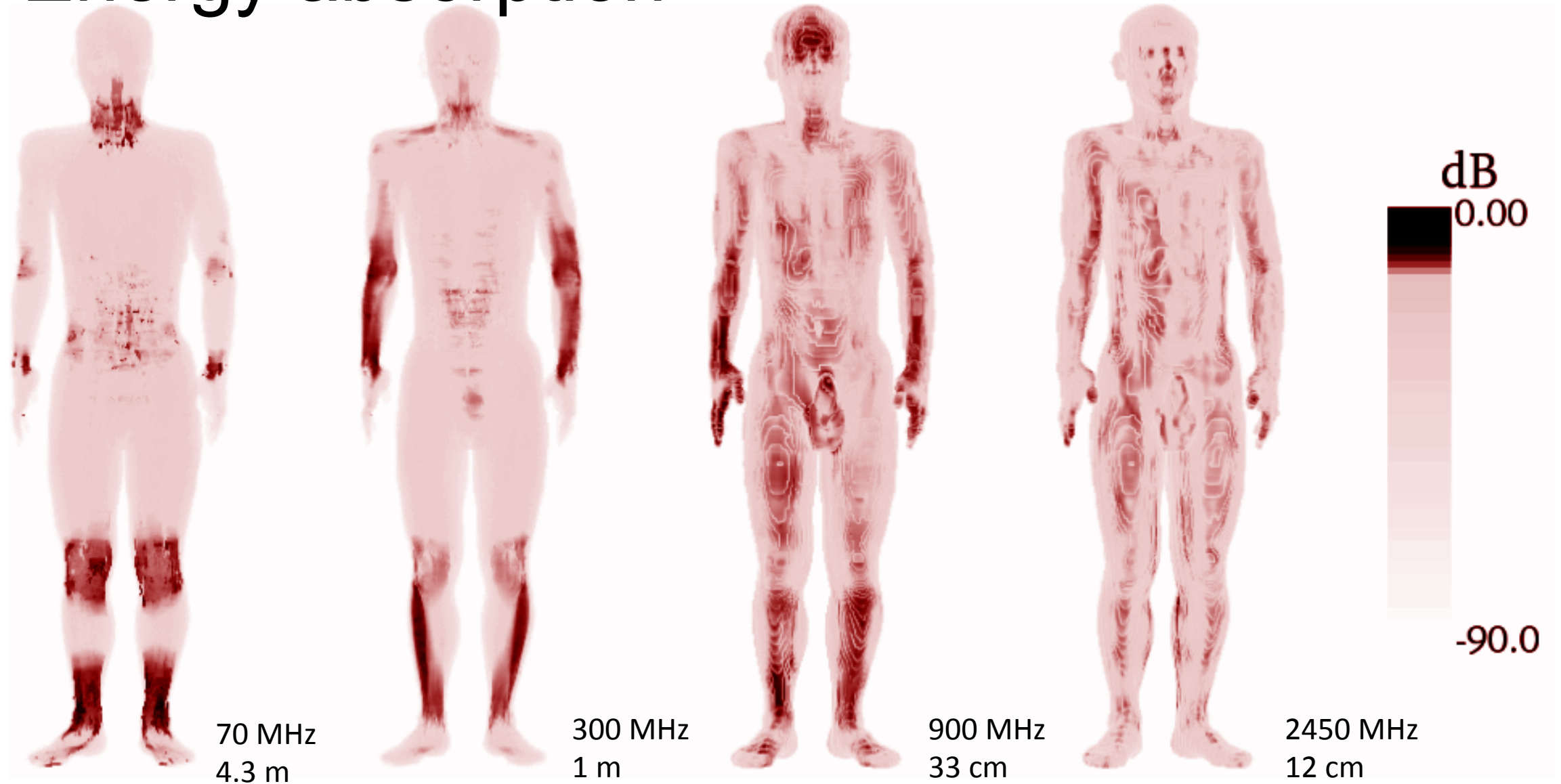
- Radar cross section and scattering
- Metamaterials
- Antenna analysis – example 7a
- Electronic component design
- Electromagnetic compatibility (EMC)
- Waveguides, resonators, filters – example 7b
- Human exposure to EM waves



# Human exposure to EM wav

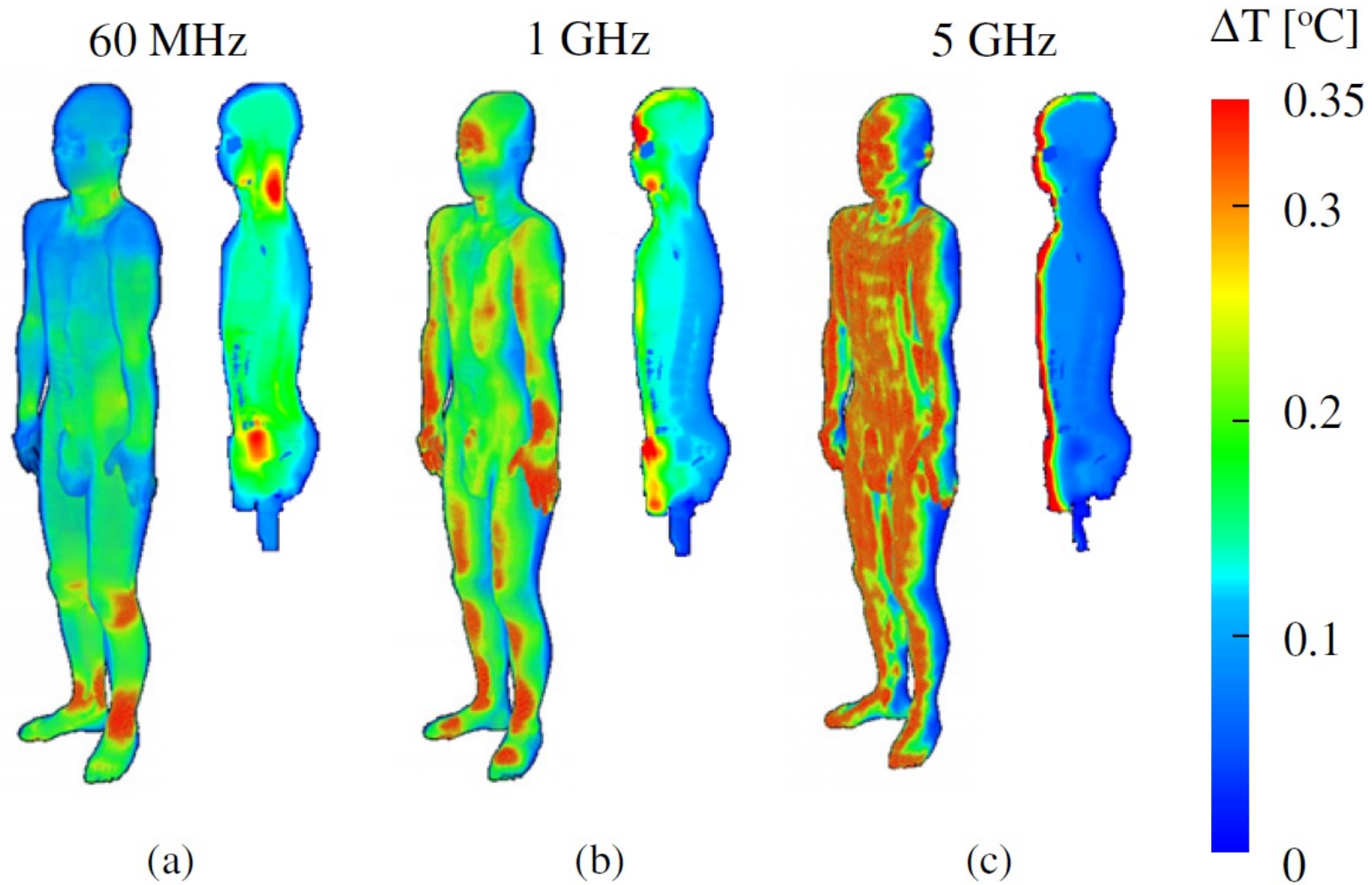


# Energy absorption

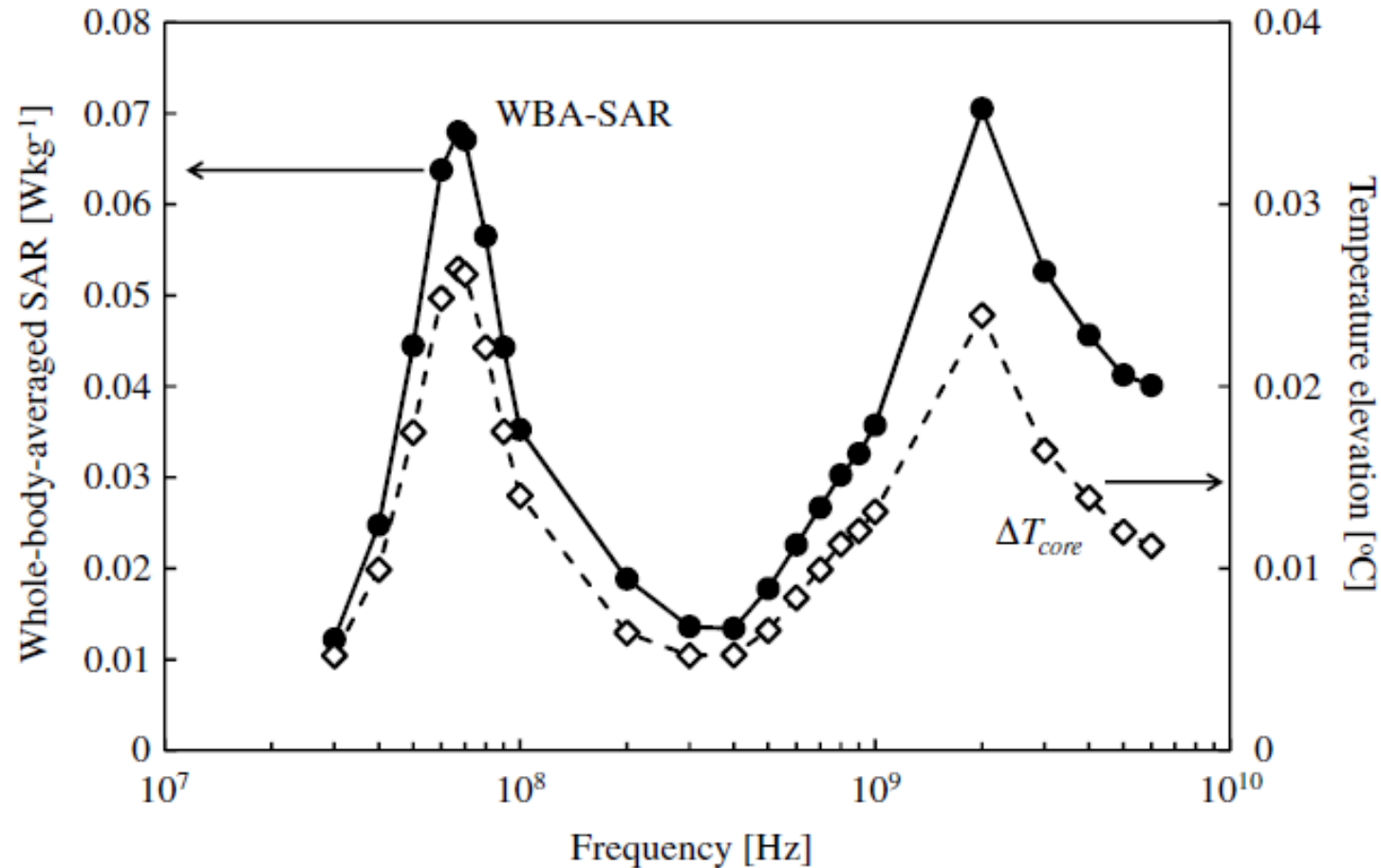


# Temperature rise

Absorbed power = 0.4 W/kg



# Power absorption versus temperature rise



# When to use FDTD?

- Use FDTD first
  - Use FDTD for making animations
  - Use FDTD for large heterogeneous geometries
  - Use FDTD to model many things simultaneously
- 
- Don't use FDTD at low frequencies
  - Don't use FDTD with too large cell size
  - Don't use FDTD if you need 99.9% accuracy