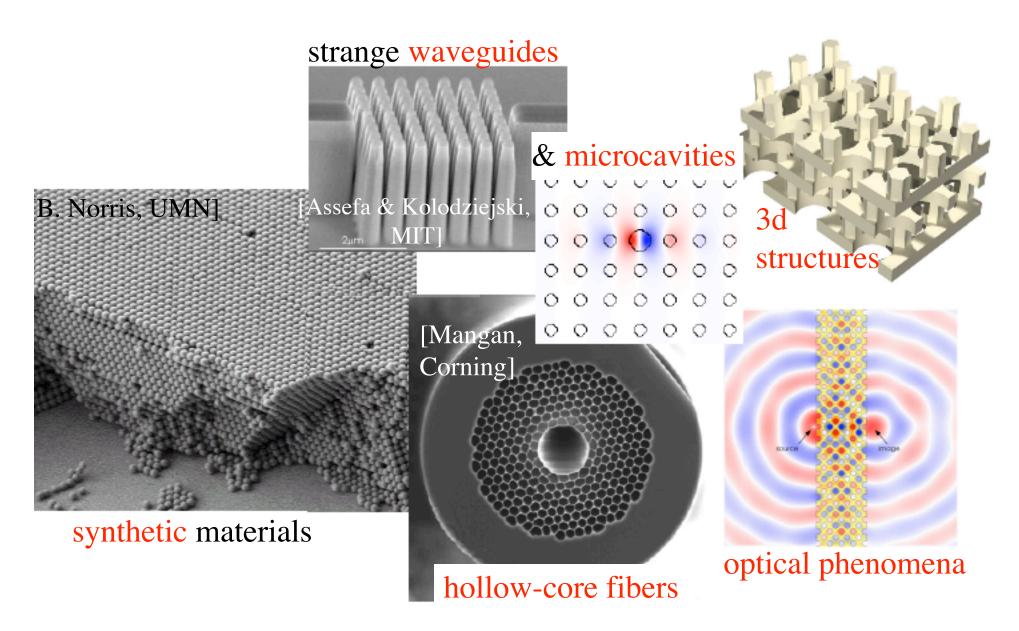
# Computational Photonics: Frequency and Time Domain Methods

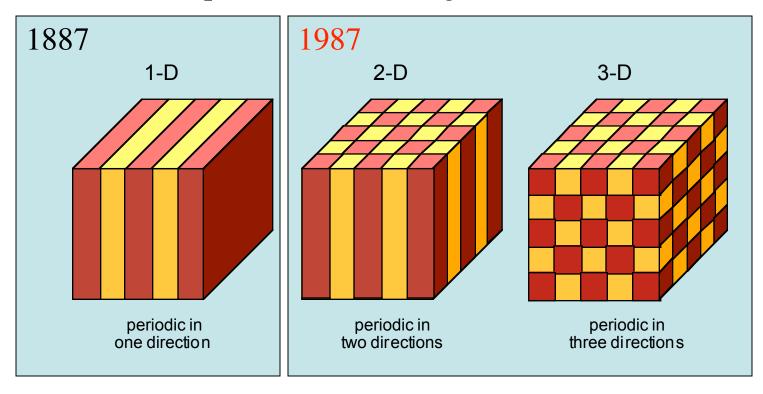
Steven G. Johnson MIT Applied Mathematics

# Nano-photonic media ( -scale)



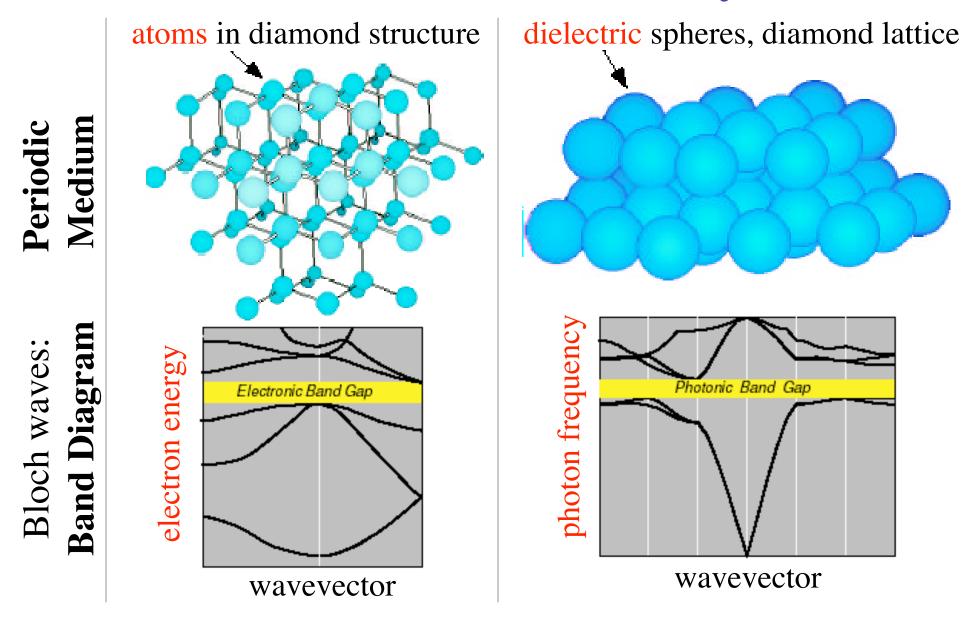
# Photonic Crystals

periodic electromagnetic media



can have a band gap: optical "insulators"

## Electronic and Photonic Crystals



# Electronic & Photonic Modeling

## Electronic

- strongly interacting
  - —entanglement, Coulomb
  - —tricky approximations

• lengthscale dependent (from Planck's *h*)

## Photonic

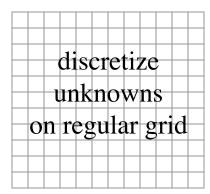
- non-interacting (or weakly),
  - simple approximations(finite resolution)
  - —any desired accuracy
- scale-invariant
  - -e.g. size  $\boxed{10}$   $\boxed{\phantom{0}}$   $\boxed{10}$  (except materials may change)

# Computational Photonics Problems

- Time-domain simulation
  - start with current  $\mathbf{J}(\mathbf{x},t)$
  - run "numerical experiment" to simulate  $\mathbf{E}(\mathbf{x}, t)$ ,  $\mathbf{H}(\mathbf{x}, t)$
- Frequency-domain linear response
  - start with harmonic current  $\mathbf{J}(\mathbf{x}, t) = e^{-i \Box t} \mathbf{J}(\mathbf{x})$
  - solve for steady-state harmonic fields E(x), H(x)
  - involves solving linear equation Ax=b
- Frequency-domain eigensolver
  - solve for source-free harmonic eigenfields  $\mathbf{E}(\mathbf{x})$ ,  $\mathbf{H}(\mathbf{x}) \sim e^{-i\Box t}$
  - involves solving eigenequation  $Ax = \square^2 x$

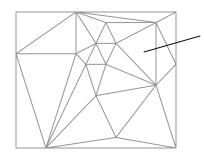
## Numerical Methods: Basis Choices

## finite difference



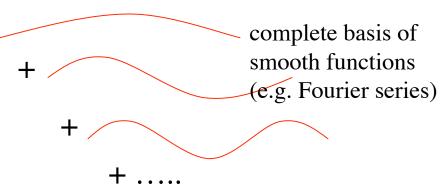
$$\frac{df}{dx} \prod \frac{f(x + ||x|) \prod f(x || ||x|)}{||x|} + O(||x^2|)$$

## finite elements

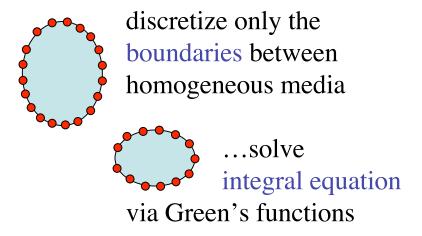


in irregular "elements," approximate unknowns by low-degree polynomial

## spectral methods

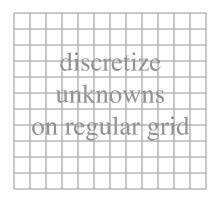


## boundary-element methods

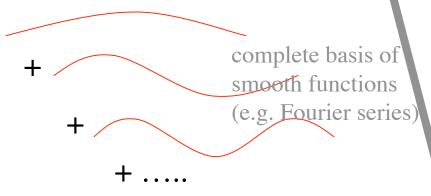


## Numerical Methods: Basis Choices

## finite difference

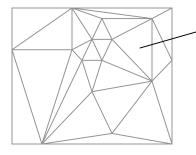


## spectral methods



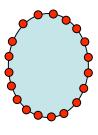
Much easier to analyze, implement, generalize, parallelize, optimize, ...

## finite elements

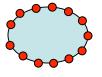


in irregular "elements," approximate unknowns by low-degree polynomial

## boundary-element methods



discretize only the boundaries between homogeneous media



...solve integral equation via Green's functions

Potentially much more efficient, especially for high resolution

## **Computational Photonics Problems**

### • Time-domain simulation

- start with current  $\mathbf{J}(\mathbf{x},t)$
- run "numerical experiment" to simulate  $\mathbf{E}(\mathbf{x}, t)$ ,  $\mathbf{H}(\mathbf{x}, t)$

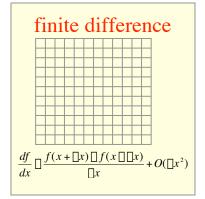
### • Frequency-domain linear response

- start with harmonic current  $\mathbf{J}(\mathbf{x}, t) = e^{-i\Box t} \mathbf{J}(\mathbf{x})$
- solve for steady-state harmonic fields E(x), H(x)
- involves solving linear equation Ax=b

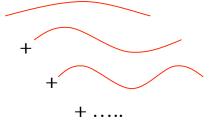
### • Frequency-domain eigensolver

- solve for source-free harmonic eigenfields  $\mathbf{E}(\mathbf{x}), \mathbf{H}(\mathbf{x}) \sim e^{-i\Box t}$
- involves solving eigenequation  $Ax = \square^2 x$

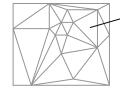
### Numerical Methods: Basis Choices



### spectral methods

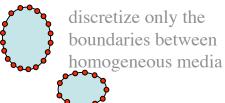


#### finite elements



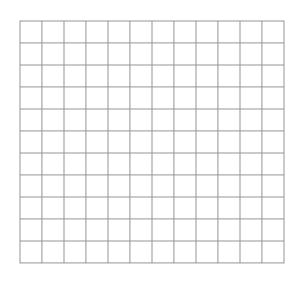
in irregular "elements," approximate unknowns by low-degree polynomial

### boundary-element methods



## **FDTD**

## Finite-Difference Time-Domain methods



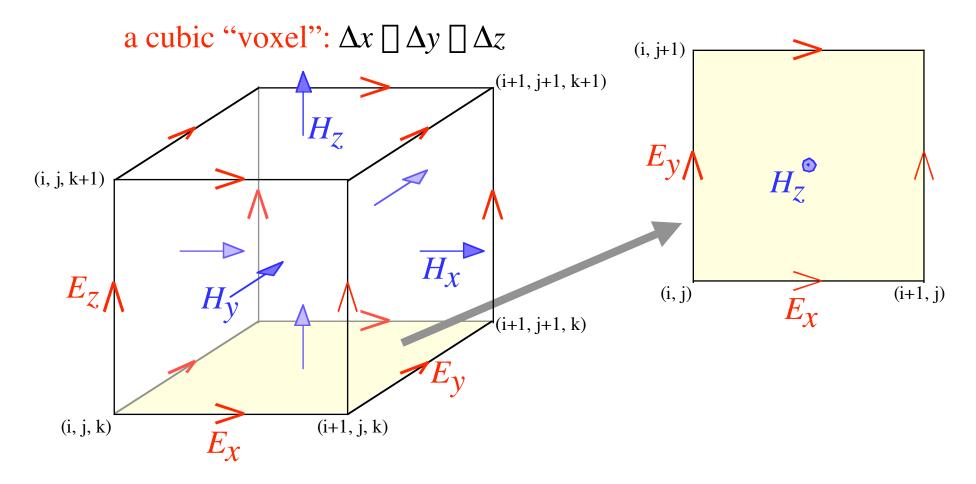
Divide both space and time into discrete grids

- spatial resolution  $\Delta x$
- temporal resolution  $\Delta t$

Very general: arbitrary geometries, materials, nonlinearities, dispersion, sources, ...

— any photonics calculation, in principle

## The Yee Discretization (1966)



Staggered grid in space:

every field component is stored on a different grid

## The Yee Discretization (1966)

$$\frac{\partial \mathbf{H}}{\partial t} = \left[ \frac{1}{\Box} \right] \mathbf{E}$$

$$\frac{\partial H_z}{\partial t} \Big|_{i+\frac{1}{2},j+\frac{1}{2}} = \left[ \frac{1}{\Box} \right] \frac{\partial E_y}{\partial x} \left[ \frac{\partial E_x}{\partial y} \right]$$

$$\frac{\partial E_z}{\partial t} \Big|_{i+\frac{1}{2},j+\frac{1}{2}} = \left[ \frac{1}{\Box} \right] \frac{\partial E_y}{\partial x} \left[ \frac{\partial E_x}{\partial y} \right]$$

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$$\frac{\partial E_x}{\partial t} \Big|_{i+\frac{1}{2},j+\frac{1}{2}} = \left[ \frac{\partial E_x}{\partial x} \right] \frac{\partial E_x}{\partial y} \Big|_{i+\frac{1}{2},j+\frac{1}{2}}$$

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$$\frac{\partial E_x}{\partial t} \Big|_{i+\frac{1}{2},j+\frac{1}{2}} = \left[ \frac{\partial E_x}{\partial x} \right] \frac{\partial E_x}{\partial x} \Big|_{i+\frac{1}{2},j+\frac{1}{2}} \Big$$

all derivatives become center differences...

## The Yee Discretization (1966)

all derivatives become *center differences*... including derivatives in *time* 

$$\frac{\partial \mathbf{H}}{\partial t}\bigg|_{t=n | t} = \frac{1}{n} \mathbf{E} \mathbf{E} \mathbf{E} = \frac{\mathbf{H}(n+\frac{1}{2}) \mathbf{H}(n | \frac{1}{2})}{\mathbf{t}} + \mathbf{O}(\Delta t^2)$$

Explicit time-stepping:

stability requires 
$$\Box t < \frac{\Box x}{\sqrt{\# \text{ dimensions}}}$$

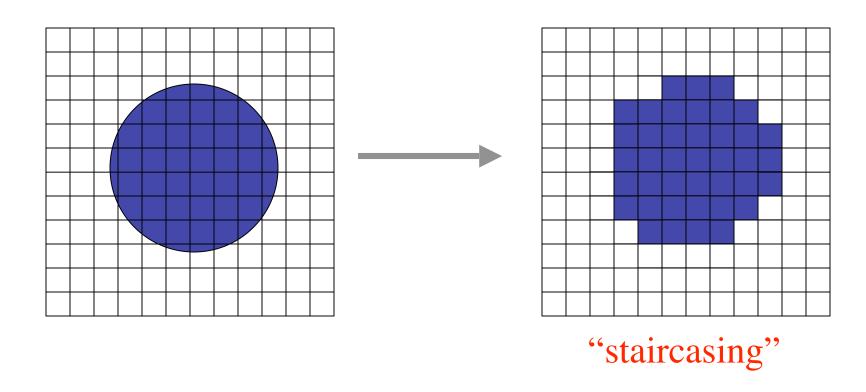
(vs. *implicit* time steps: invert large matrix at each step)

# FDTD Discretization Upshot

- For stability, space and time resolutions are proportional
  - doubling resolution in 3d requires at least  $16 = 2^4$  times the work!
- But at least the error goes quadratically with resolution ...right?

...not necessarily!

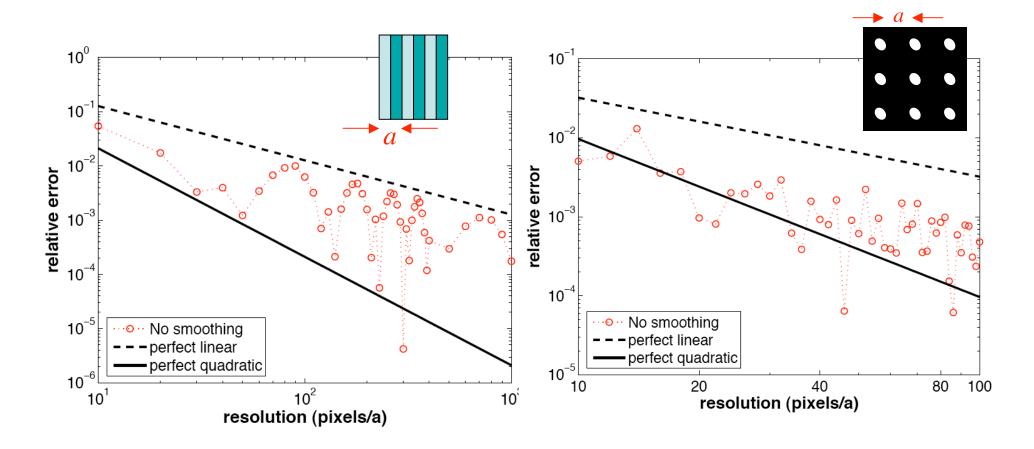
# Difficulty with a grid: representing discontinuous materials?



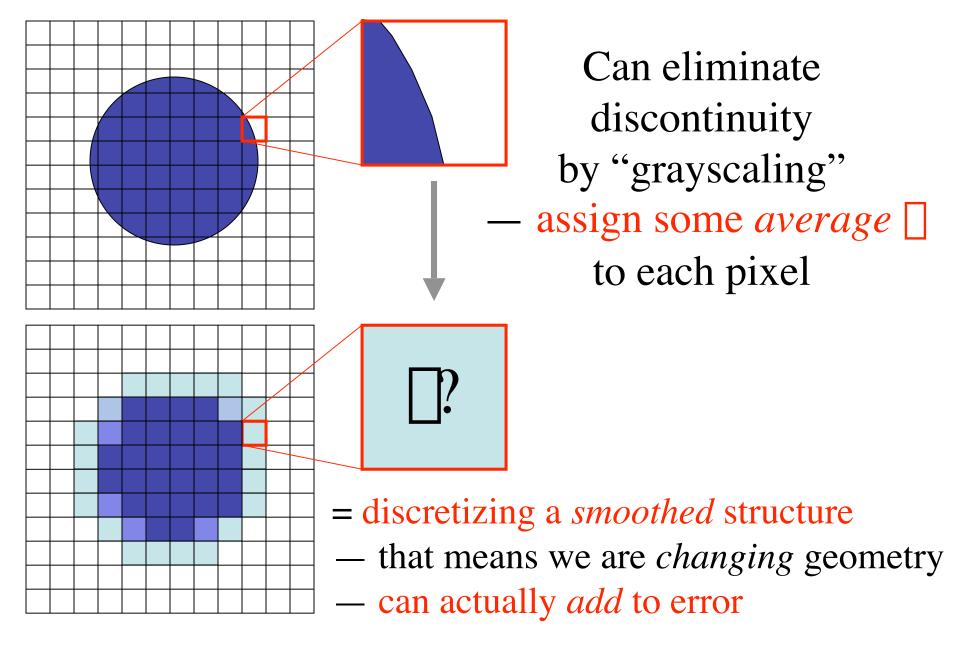
... how does this affect accuracy?

# Field Discontinuity Degrades Order of Accuracy

TE polarization (E in plane: discontinuous)

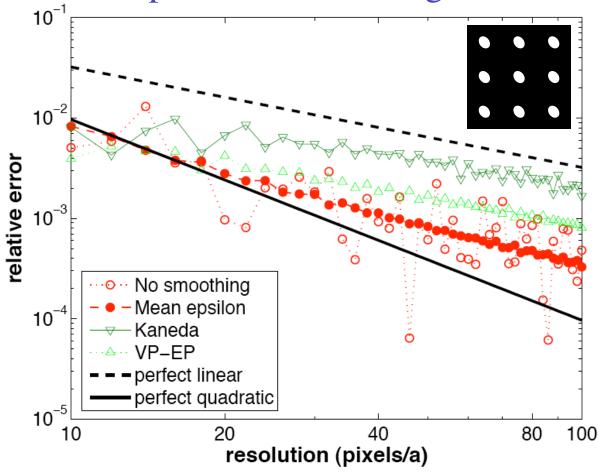


# Sub-pixel smoothing



# Past sub-pixel smoothing methods can make error worse!



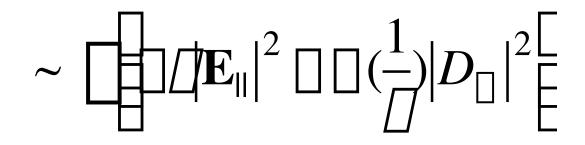


& convergence is still only linear

[ Dey, 1999 ] [ Kaneda, 1997 ] [ Mohammadi, 2005]

# A Criterion for Accurate Smoothing

1st-order errors from smoothing []

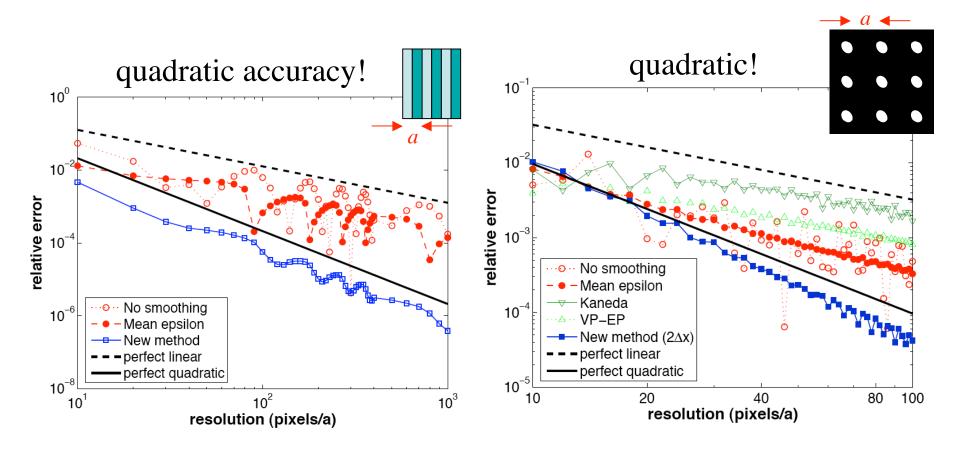


We want the smoothing errors to be zero to 1st order

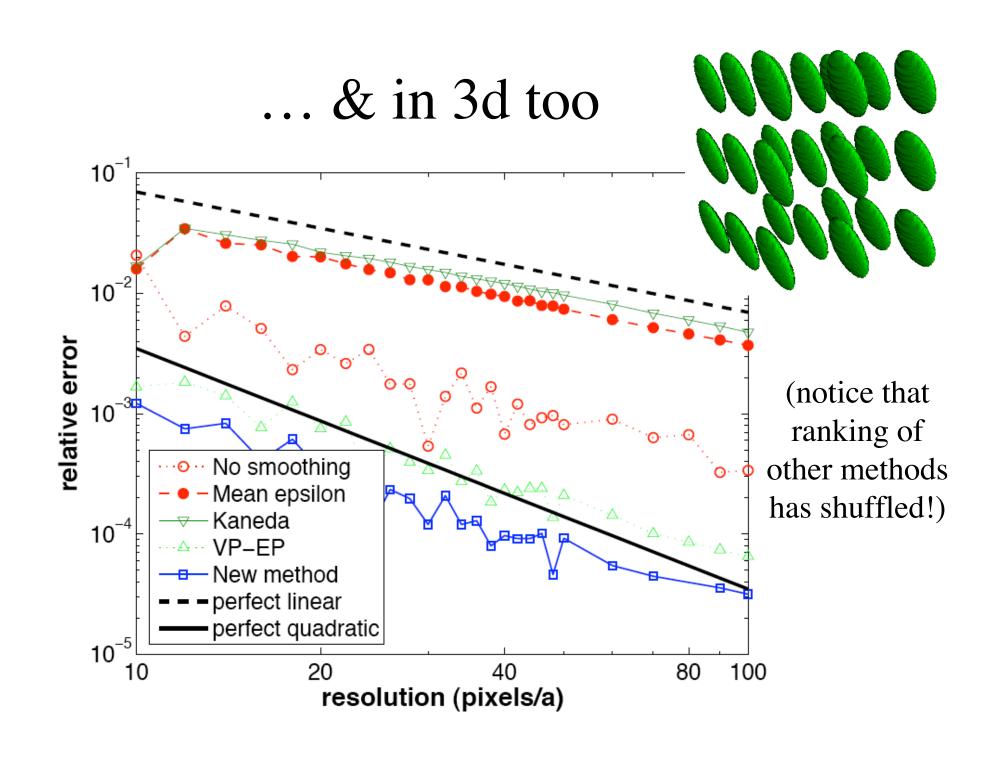
— minimizes error and 2nd-order convergent!

$$\begin{array}{c|c} & Use \ a \ tensor \ \hline \\ & & \\ &$$

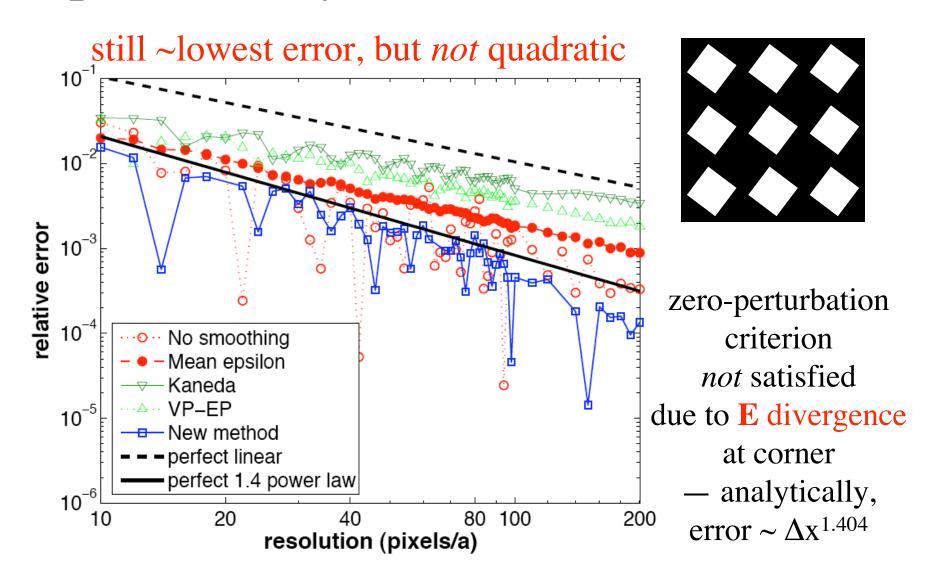
# Consistently the Lowest Error



[ Farjadpour et al., Opt. Lett. 2006 ]



# A qualitatively different case: corners

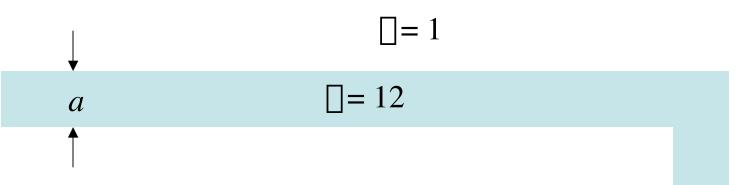


# Yes, but what can you do with FDTD?

### Some common tasks:

- Frequency-domain response:
  - put in harmonic source and wait for steady-state
- Transmission/reflection spectra:
  - get entire spectrum from a single simulation
     (Fourier transform of impulse response)
- Eigenmodes and resonant modes:
  - get all modes from a single simulation (some tricky signal processing)

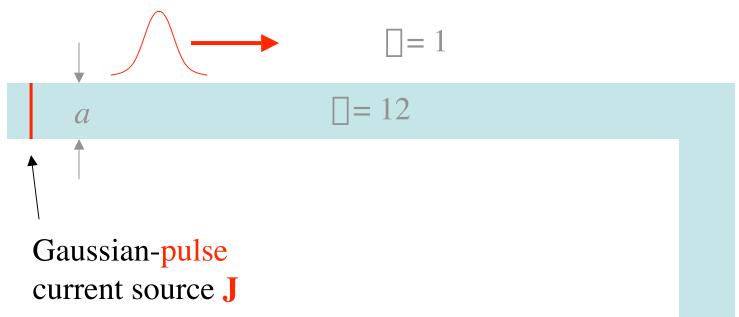
# Transmission Spectra in FDTD



example: a 90° bend, 2d strip waveguide

transmitted power = energy flux here:

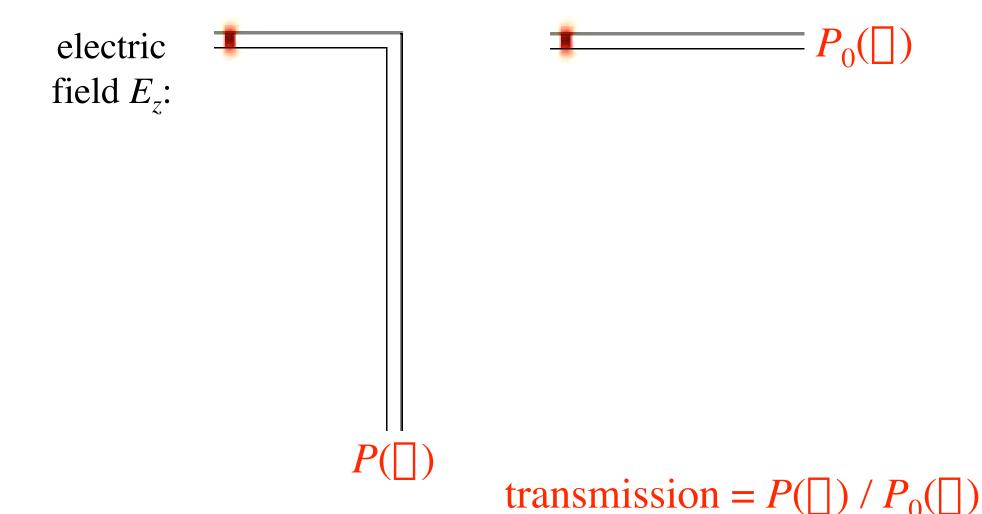
# Transmission Spectra in FDTD



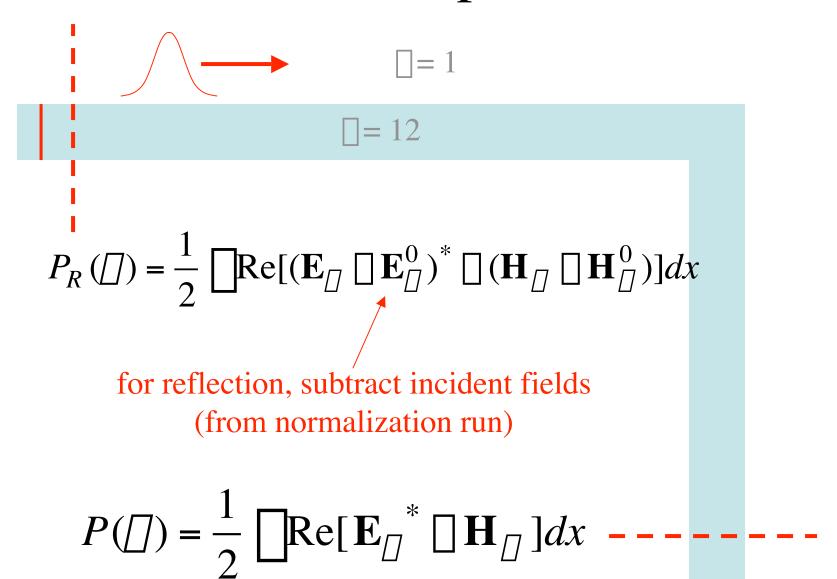
Fourier-transform the fields at each x:

# Transmission Spectra in FDTD

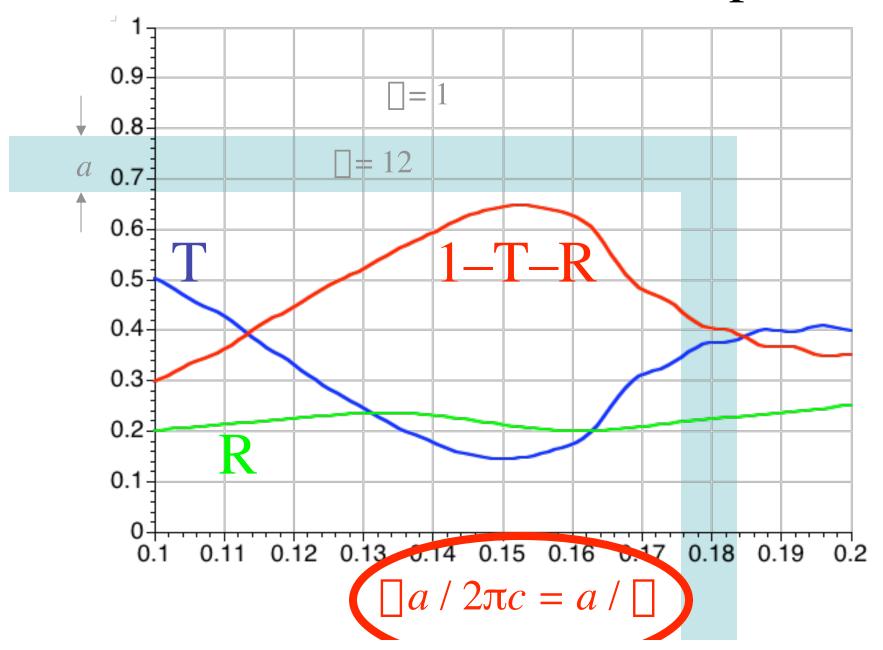
must always do two simulations: one for normalization



# Reflection Spectra in FDTD



# Transmission/Reflection Spectra



## **Dimensionless Units**

## Maxwell's equations are scale invariant

- most useful quantities are dimensionless ratios like a /  $\square$ , for a characteristic lengthscale a
- same ratio, same  $\square$  = same solution regardless of whether  $a = 1\mu$ m or 1km

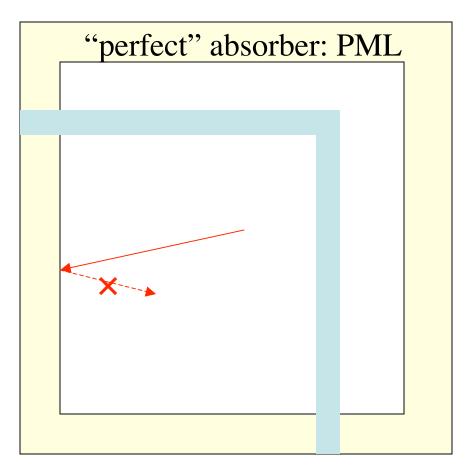
## Our (typical) approach:

pick characteristic lengthscale a

- measure distance in units of a
- measure time in units of a/c
- measure  $\square$  in units of  $2\pi c/a = a / \square$

**—** ....

# Absorbing Boundaries: Perfectly Matched Layers



Artificial absorbing material *overlapping* the computation

Theoretically reflectionless

... but PML is no longer perfect with finite resolution, so "gradually turn on" absorption over finite-thickness PML

## Computational Photonics Problems

### • Time-domain simulation

- start with current  $J(\mathbf{x},t)$
- run "numerical experiment" to simulate  $\mathbf{E}(\mathbf{x}, t)$ ,  $\mathbf{H}(\mathbf{x}, t)$

### • Frequency-domain linear response

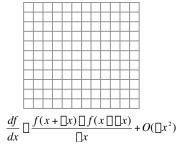
- start with harmonic current  $\mathbf{J}(\mathbf{x}, t) = e^{-i\Box t} \mathbf{J}(\mathbf{x})$
- solve for steady-state harmonic fields E(x), H(x)
- involves solving linear equation Ax=b

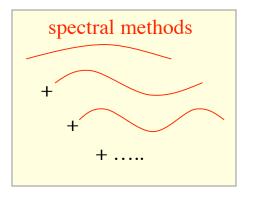
### • Frequency-domain eigensolver

- solve for source-free harmonic eigenfields  $\mathbf{E}(\mathbf{x}), \mathbf{H}(\mathbf{x}) \sim e^{-i\Box t}$
- involves solving eigenequation  $Ax = \square^2 x$

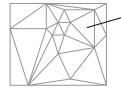
### Numerical Methods: Basis Choices

#### finite difference



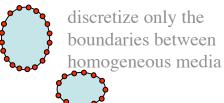


#### finite elements



in irregular "elements," approximate unknowns by low-degree polynomial

### boundary-element methods



# A Maxwell Eigenproblem

$$\vec{\Box} \vec{E} = \vec{\Box} \frac{1}{c} \frac{\partial}{\partial t} \vec{H} = i \frac{\vec{\Box}}{c} \vec{H}$$

$$\vec{\Box} \vec{D} \vec{H} = \vec{\Box} \frac{1}{c} \frac{\partial}{\partial t} \vec{E} + \vec{J} = \vec{\Box} i \frac{\vec{\Box}}{c} \vec{E}$$

dielectric function  $\Pi(\mathbf{x}) = n^2(\mathbf{x})$ 

First task: get rid of this mess

# Electronic & Photonic Eigenproblems

## Electronic

$$\frac{1}{2m} \frac{\hbar^2}{2m} = E$$

nonlinear eigenproblem (V depends on e density  $| / |^2$ )

## **Photonic**

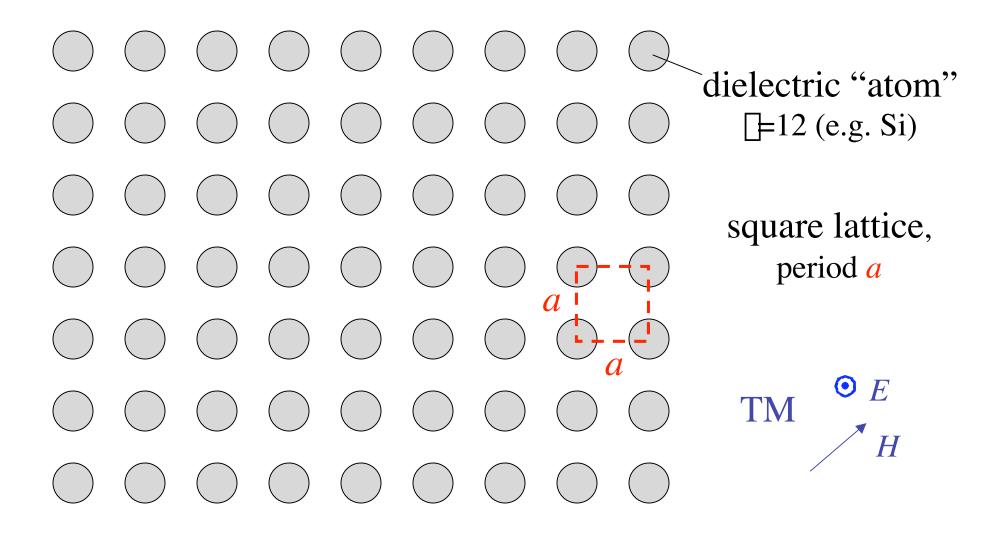
$$\frac{1}{2m} \frac{\hbar^2}{2m} \Box^2 + V \overrightarrow{D} = E \Box \qquad \Box \overrightarrow{D} \Box \overrightarrow{H} = \overrightarrow{D} \overrightarrow{H}$$

simple linear eigenproblem (for linear materials)

-many well-known computational techniques

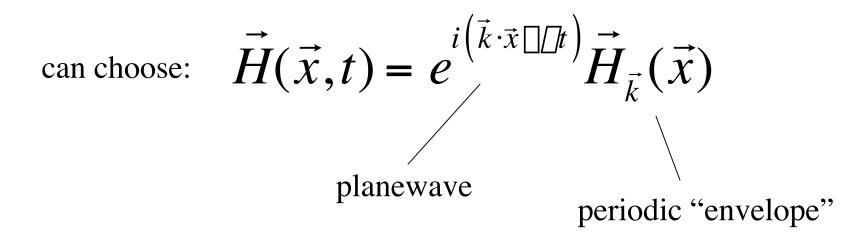
Hermitian = real E & [], ... Periodicity = Bloch's theorem...

# A 2d Model System



# Periodic Eigenproblems

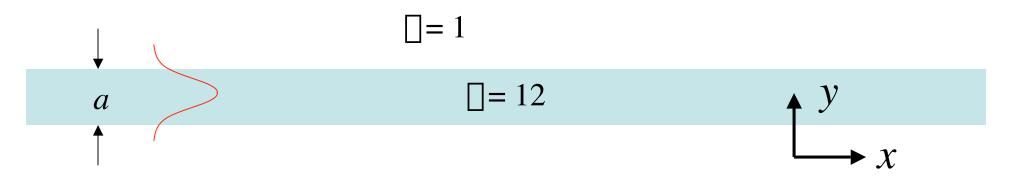
if eigen-operator is periodic, then Bloch-Floquet theorem applies:



Corollary 1: k is conserved, i.e. no scattering of Bloch wave

Corollary 2: 
$$\vec{H}_{\vec{k}}$$
 given by finite unit cell, so  $\square$  are discrete  $\square_n(\mathbf{k})$ 

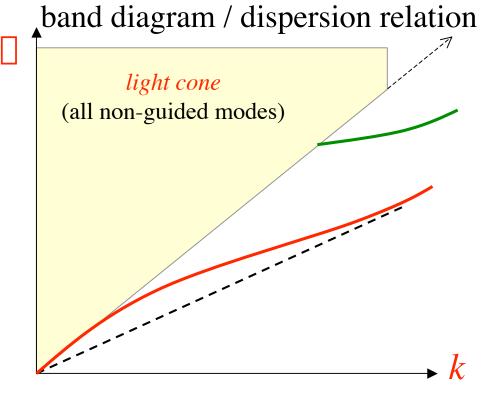
# A More Familiar Eigenproblem



find the normal modes of the waveguide:

$$\mathbf{H}(y,t) = \mathbf{H}_k(y)e^{i(kx\square t)}$$

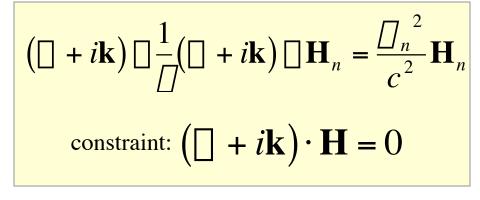
(propagation constant k  $a.k.a. \square$ )

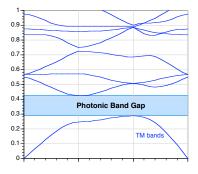


## Solving the Maxwell Eigenproblem

*Finite* cell  $\rightarrow$  *discrete* eigenvalues  $\square_n$ 

Want to solve for  $\square_n(\mathbf{k})$ , & plot vs. "all" **k** for "all" n,



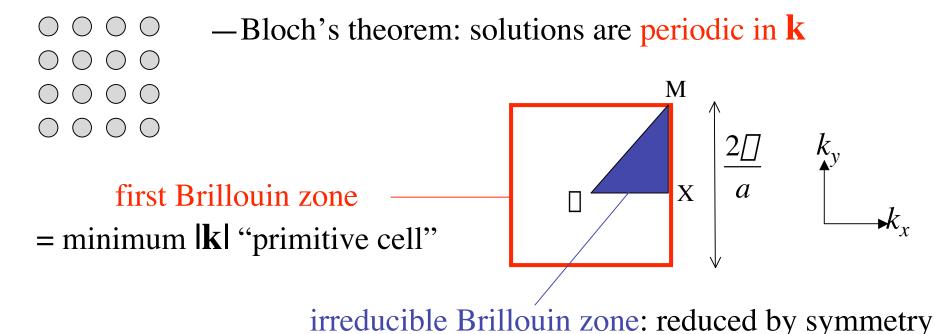


where:  $\mathbf{H}(x,y) \mathbf{\Box}^{i(\mathbf{k}\cdot\mathbf{x} - \Box t)}$ 

- 1 Limit range of  $\mathbf{k}$ : irreducible Brillouin zone
- 2 Limit degrees of freedom: expand **H** in finite basis
- 3 Efficiently solve eigenproblem: iterative methods

## Solving the Maxwell Eigenproblem: 1

 $\mathbf{1}$  Limit range of  $\mathbf{k}$ : irreducible Brillouin zone



- 2 Limit degrees of freedom: expand **H** in finite basis
- 3 Efficiently solve eigenproblem: iterative methods

## Solving the Maxwell Eigenproblem: 2a

- 1 Limit range of **k**: irreducible Brillouin zone
- 2 Limit degrees of freedom: expand **H** in finite basis (N)

$$|\mathbf{H}\rangle = \mathbf{H}(\mathbf{x}_t) = \prod_{m=1}^{N} h_m \mathbf{b}_m(\mathbf{x}_t)$$
 solve:  $\hat{A}|\mathbf{H}\rangle = \prod^{2} |\mathbf{H}\rangle$ 

finite matrix problem:  $Ah = \prod^2 Bh$ 

$$\langle \mathbf{f} | \mathbf{g} \rangle = \mathbf{f}^* \cdot \mathbf{g}$$
  $A_{m\ell} = \langle \mathbf{b}_m | \hat{A} | \mathbf{b}_\ell \rangle$   $B_{m\ell} = \langle \mathbf{b}_m | \mathbf{b}_\ell \rangle$ 

3 Efficiently solve eigenproblem: iterative methods

## Solving the Maxwell Eigenproblem: 2b

- 1 Limit range of **k**: irreducible Brillouin zone
- 2 Limit degrees of freedom: expand **H** in finite basis

— must satisfy constraint:  $(\Box + i\mathbf{k}) \cdot \mathbf{H} = 0$ 

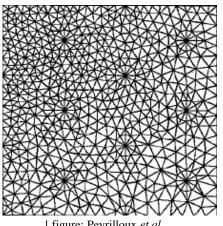
#### Planewave (FFT) basis

# $\mathbf{H}(\mathbf{x}_t) = \prod_{\mathbf{G}} \mathbf{H}_{\mathbf{G}} e^{i\mathbf{G} \cdot \mathbf{x}_t}$

constraint: 
$$\mathbf{H}_{\mathbf{G}} \cdot (\mathbf{G} + \mathbf{k}) = 0$$

uniform "grid," periodic boundaries, simple code, O(N log N)

#### Finite-element basis



[ figure: Peyrilloux *et al.*, *J. Lightwave Tech*. **21**, 536 (2003) ]

constraint, boundary conditions:

#### Nédélec elements

[ Nédélec, *Numerische Math.* **35**, 315 (1980) ]

nonuniform mesh, more arbitrary boundaries, complex code & mesh, O(N)

3 Efficiently solve eigenproblem: iterative methods

## Solving the Maxwell Eigenproblem: 3a

- 1 Limit range of **k**: irreducible Brillouin zone
- 2 Limit degrees of freedom: expand **H** in finite basis
- 3 Efficiently solve eigenproblem: iterative methods

$$Ah = \prod^2 Bh$$

Slow way: compute A & B, ask LAPACK for eigenvalues

— requires  $O(N^2)$  storage,  $O(N^3)$  time

#### Faster way:

- start with *initial guess* eigenvector  $h_0$
- *iteratively* improve
- O(Np) storage, ~  $O(Np^2)$  time for p eigenvectors (p smallest eigenvalues)

## Solving the Maxwell Eigenproblem: 3b

- 1 Limit range of **k**: irreducible Brillouin zone
- 2 Limit degrees of freedom: expand **H** in finite basis
- 3 Efficiently solve eigenproblem: iterative methods

$$Ah = \prod^2 Bh$$

#### Many iterative methods:

Arnoldi, Lanczos, Davidson, Jacobi-Davidson, ...,
 Rayleigh-quotient minimization

## Solving the Maxwell Eigenproblem: 3c

- 1 Limit range of **k**: irreducible Brillouin zone
- 2 Limit degrees of freedom: expand **H** in finite basis
- 3 Efficiently solve eigenproblem: iterative methods

$$Ah = \prod^2 Bh$$

Many iterative methods:

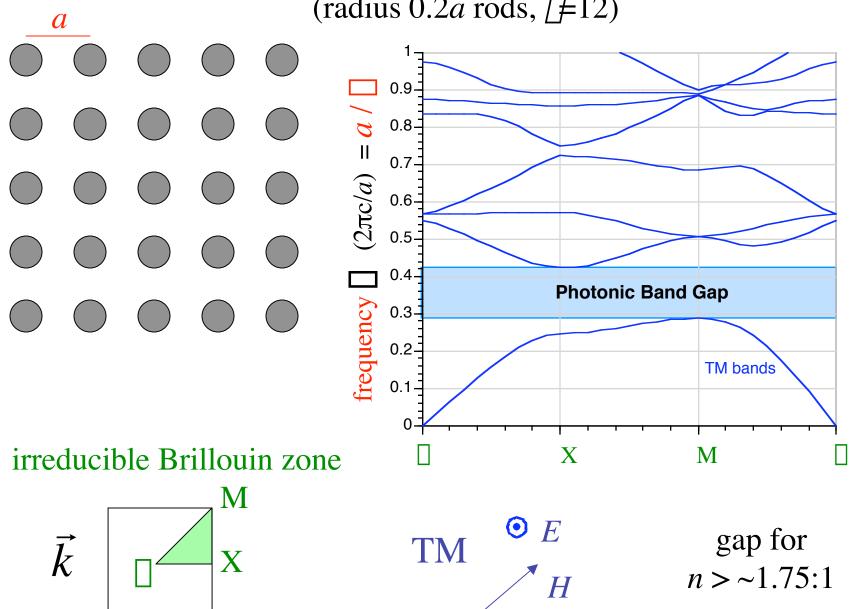
Arnoldi, Lanczos, Davidson, Jacobi-Davidson, ...,
 Rayleigh-quotient minimization

for Hermitian matrices, smallest eigenvalue  $\square_0$  minimizes:

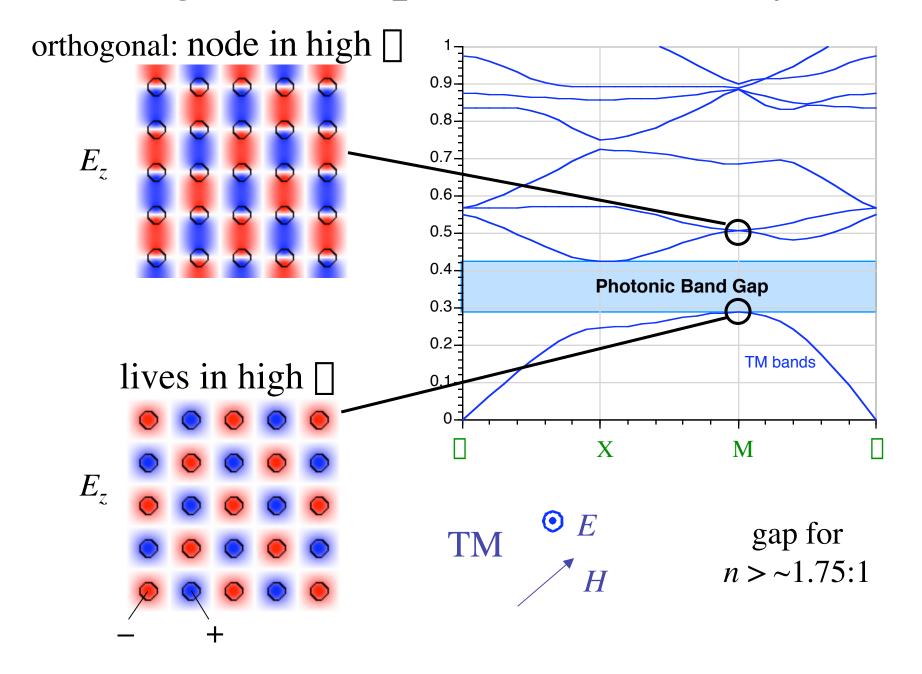
minimize by preconditioned conjugate-gradient (or...)

#### Band Diagram of 2d Model System

(radius 0.2a rods,  $\not\equiv 12$ )



## Origin of Gap in 2d Model System



### The Iteration Scheme is Important

(minimizing function of  $10^4$ – $10^8$ + variables!)

**Steepest-descent:** minimize  $(h + \square \square f)$  over  $\square$  ... repeat

Conjugate-gradient: minimize  $(h + \square d)$ 

— d is  $\prod f + (stuff)$ : conjugate to previous search dirs

Preconditioned steepest descent: minimize  $(h + \square d)$ 

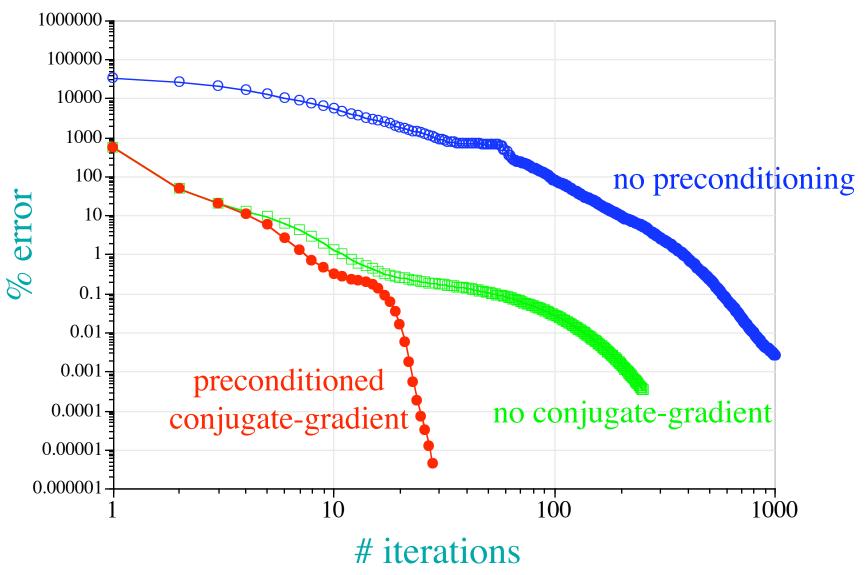
$$-d = (approximate A^{-1}) \Box f \sim Newton's method$$

Preconditioned conjugate-gradient: minimize  $(h + \square d)$ 

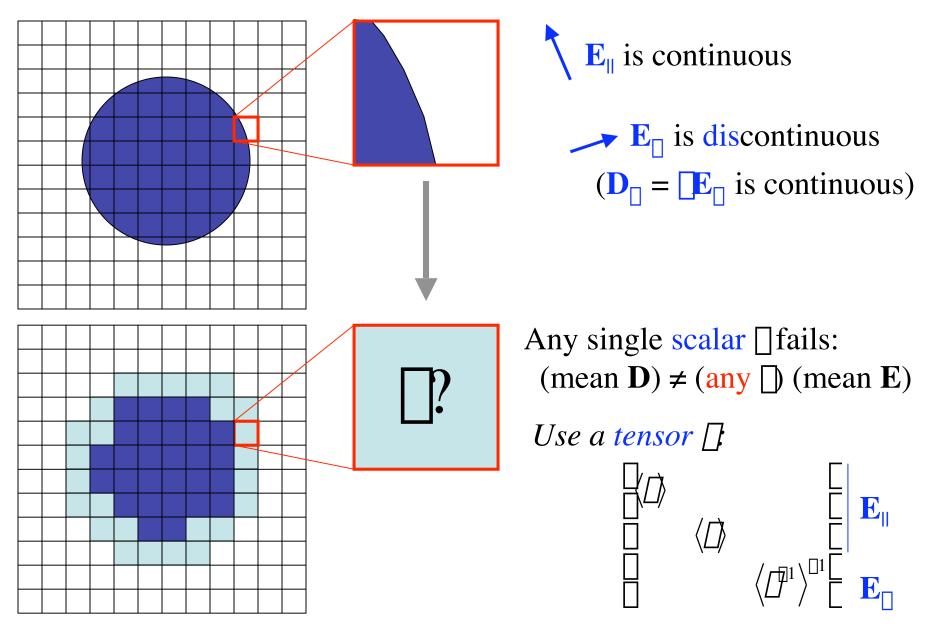
— 
$$d$$
 is (approximate A<sup>-1</sup>) [ $\Box f$  + (stuff)]

#### The Iteration Scheme is Important

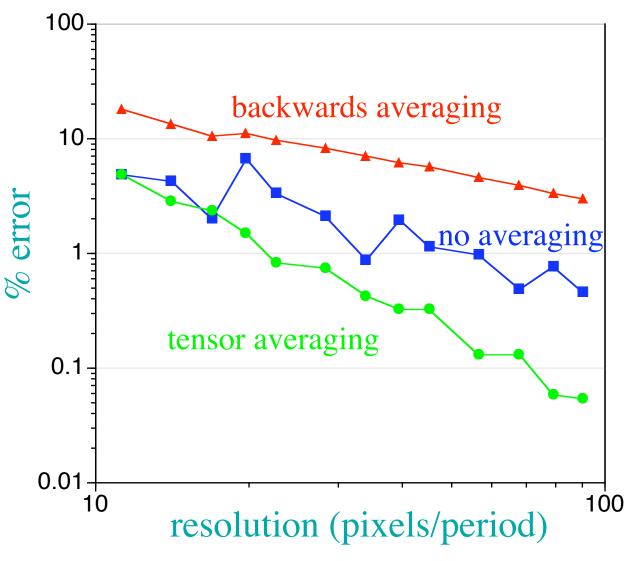
(minimizing function of ~40,000 variables)



## The Boundary Conditions are Tricky



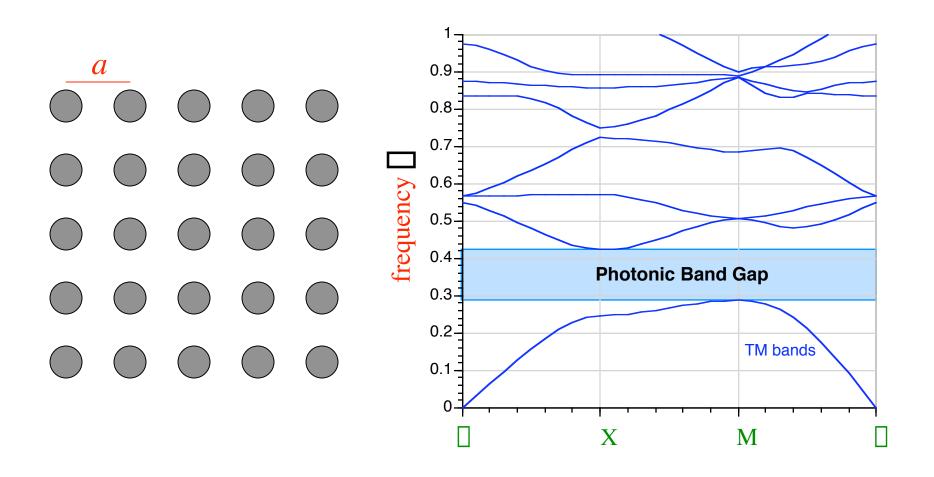
## The Daveraging is Important



changes *order*of convergence
from  $\Delta x$  to  $\Delta x^2$ 

(similar effects in other E&M numerics & analyses)

## Gap, Schmap?

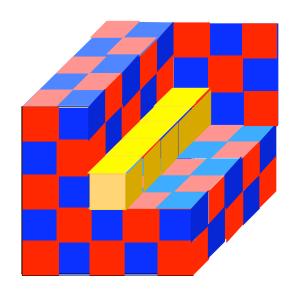


But, what can we do with the gap?

## Intentional "defects" are good

microcavities

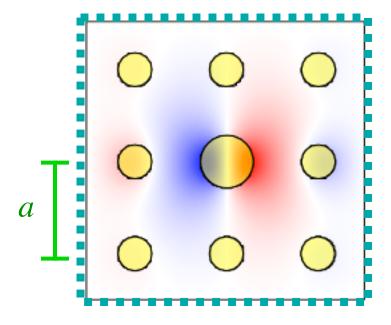
waveguides ("wires")



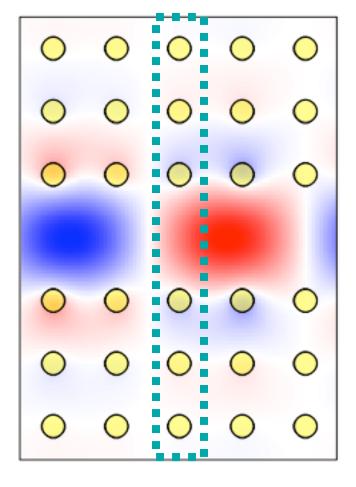
#### Intentional "defects" in 2d

(Same computation, with supercell = many primitive cells)

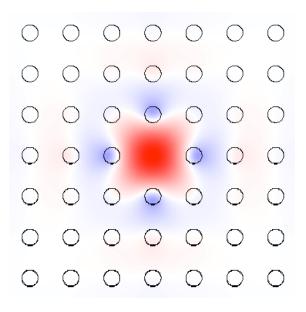
#### microcavities



#### waveguides



### Microcavity Blues

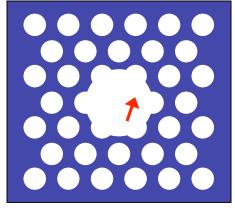


For cavities (*point defects*) frequency-domain has its drawbacks:

- Best methods compute lowest □ bands,
   but N<sup>d</sup> supercells have N<sup>d</sup> modes
   below the cavity mode expensive
- Best methods are for Hermitian operators, but losses requires non-Hermitian

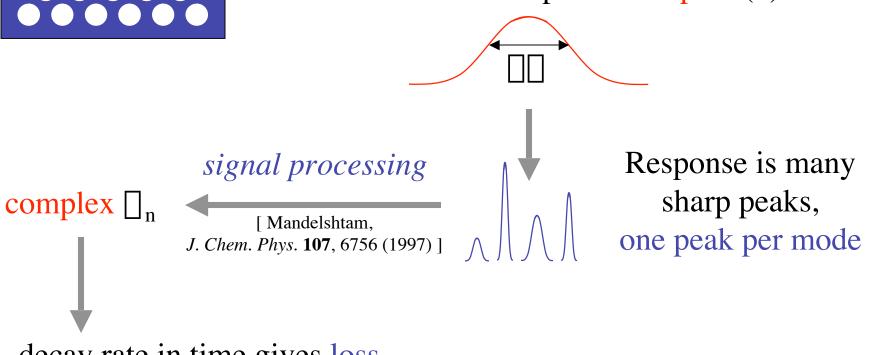
### Time-Domain Eigensolvers

(finite-difference time-domain = FDTD)



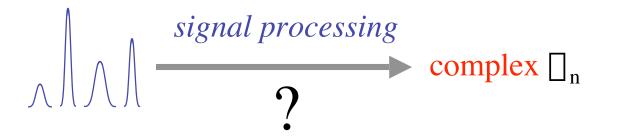
Simulate Maxwell's equations on a discrete grid, + absorbing boundaries (leakage loss)

• Excite with broad-spectrum dipole (\*) source

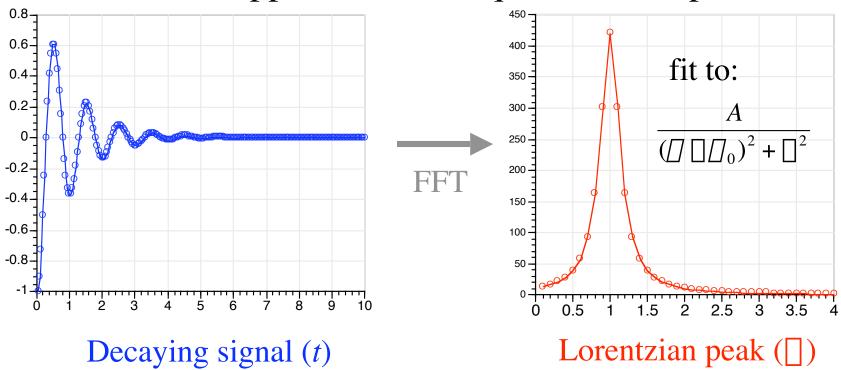


decay rate in time gives loss

## Signal Processing is Tricky

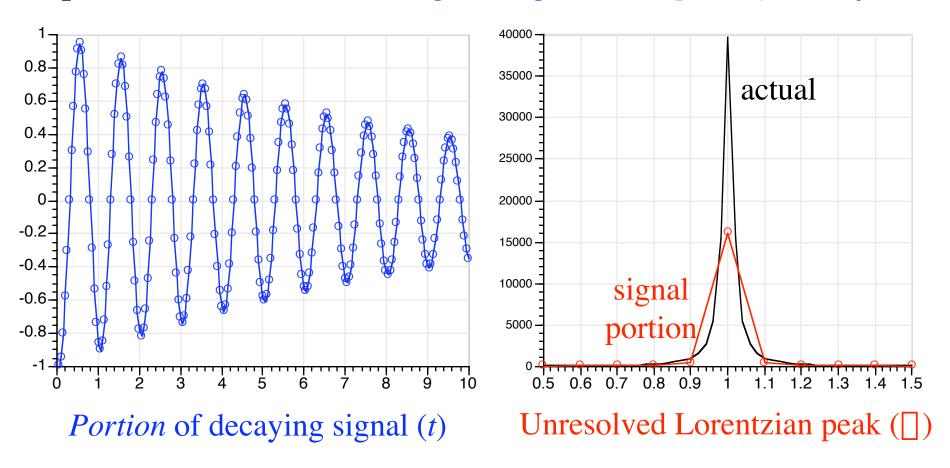


a common approach: least-squares fit of spectrum



### Fits and Uncertainty

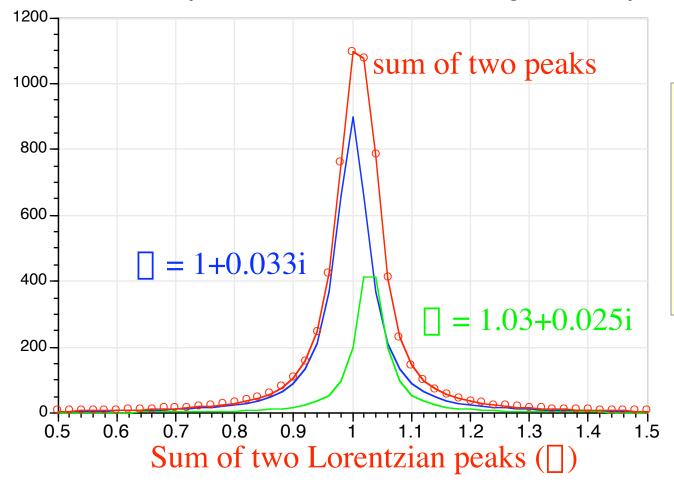
problem: have to run long enough to *completely* decay



There is a better way, which gets complex  $\square$  to > 10 digits

### Unreliability of Fitting Process

Resolving two overlapping peaks is near-impossible 6-parameter nonlinear fit (too many local minima to converge reliably)



There is a better way, which gets complex for both peaks to > 10 digits

Quantum-inspired signal processing (NMR spectroscopy):

## Filter-Diagonalization Method (FDM)

[ Mandelshtam, J. Chem. Phys. **107**, 6756 (1997) ]

Given time series 
$$y_n$$
, write:  $y_n = y(n \square t) = \square_k a_k e^{\square i \square_k n \square t}$ 

...find *complex* amplitudes  $a_k$  & frequencies  $\Box_k$  by a simple linear-algebra problem!

Idea: pretend y(t) is autocorrelation of a quantum system:

$$\hat{H}|\Box\rangle = i\hbar \frac{\partial}{\partial t}|\Box\rangle$$
 time- $\Delta t$  evolution-operator:  $\hat{U} = e^{\Box i\hat{H}\Box t/\hbar}$ 

say: 
$$y_n = \langle [0] | [n] t \rangle = \langle [0] | \hat{U}^n | [0] \rangle$$

### Filter-Diagonalization Method (FDM)

[ Mandelshtam, J. Chem. Phys. **107**, 6756 (1997) ]

$$y_n = \langle \Box(0) | \Box(n\Box t) \rangle = \langle \Box(0) | \hat{U}^n | \Box(0) \rangle \qquad \hat{U} = e^{\Box i \hat{H} \Box t / \hbar}$$

We want to diagonalize U: eigenvalues of U are  $e^{i\Box \Delta t}$  ... expand U in basis of  $|\Box(n\Delta t)\rangle$ :

$$U_{m,n} = \langle \Box(m\Box t) | \hat{U} | \Box(n\Box t) \rangle = \langle \Box(0) | \hat{U}^m \hat{U} \hat{U}^n | \Box(0) \rangle = y_{m+n+1}$$

 $U_{mn}$  given by  $y_n$ 's — just diagonalize known matrix!

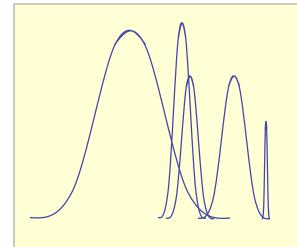
## Filter-Diagonalization Summary

[ Mandelshtam, J. Chem. Phys. 107, 6756 (1997) ]

#### $U_{mn}$ given by $y_n$ 's — just diagonalize known matrix!

A few omitted steps:

- —Generalized eigenvalue problem (basis not orthogonal)
- —Filter  $y_n$ 's (Fourier transform): small bandwidth = smaller matrix (less singular)



- resolves many peaks at once
- # peaks not known a priori
- resolve overlapping peaks
- resolution >> Fourier uncertainty

#### Do try this at home

#### **FDTD** simulation:

```
http://ab-initio.mit.edu/meep/
```

#### Bloch-mode eigensolver:

```
http://ab-initio.mit.edu/mpb/
```

#### Filter-diagonalization:

```
http://ab-initio.mit.edu/harminv/
```

```
Photonic-crystal tutorials (+ THIS TALK):
    http://ab-initio.mit.edu/
    /photons/tutorial/
```

### Meep (FDTD)

- Arbitrary □(x) including dispersive, loss/gain, and nonlinear [□(2) and □(3)]
- Arbitrary  $\mathbf{J}(\mathbf{x},t)$
- PML/periodic/metal bound.
- 1d/2d/3d/cylindrical
- power spectraeigenmodes

### MPB (Eigensolver)

- Arbitrary periodic [(x) anisotropic, magneto-optic, ... (lossless, linear materials)
  - 1d/2d/3d
- band diagrams, group velocities perturbation theory, ...



- MPI parallelism
- exploit mirror symmetries

- fully scriptable interface
- built-in multivariate optimization, integration, root-finding, ...
- field output (standard HDF5 format)

### Unix Philosophy

combine small, well-designed tools, via files

Input text file  $\longrightarrow$  MPB/Meep  $\longrightarrow$  standard formats (text + HDF5)

#### Disadvantage:

have to learn several programs

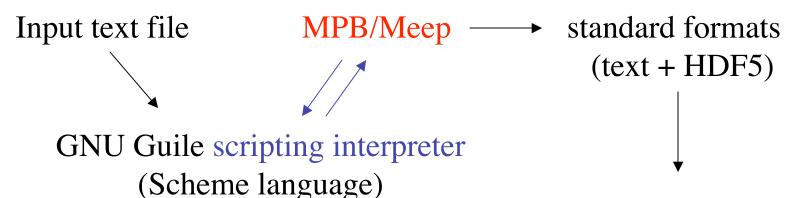
#### Advantages:

- flexibility
- batch processing, shell scripting
- ease of development

Visualization / Analysis software (Matlab, Mayavi [vtk], command-line tools, ...)

### Unix Philosophy

combine small, well-designed tools, via files

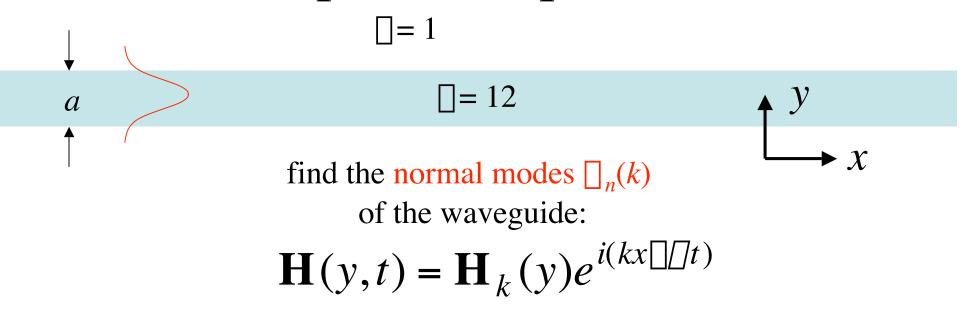


Embed a full scripting language:

- parameter sweeps
- complex parameterized geometries
- optimization, integration, etc.
- programmable J(x, t), etc.
- ... Turing complete

Visualization / Analysis software (Matlab, Mayavi [vtk], command-line tools, ...)

### A Simple Example (MPB)



#### Need to specify:

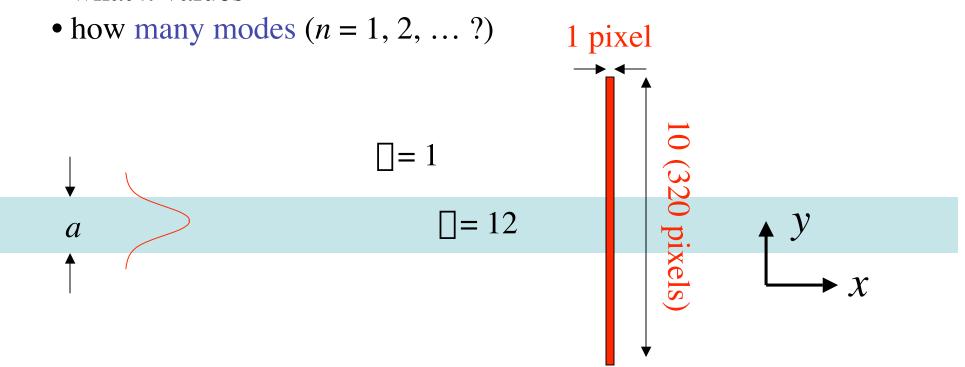
- computational cell size/resolution
- geometry, i.e.  $\boxed{y}$
- what *k* values
- how many modes  $(n = 1, 2, \dots ?)$

#### Need to specify:

• computational cell size/resolution

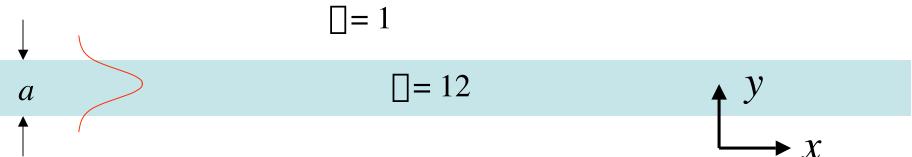
```
(set! geometry-lattice (make lattice (size no-size 10 no-size)
(set! resolution 32)
```

- geometry, i.e. y
- what *k* values



#### Need to specify:

```
• computational cell size/resolution
```



#### Need to specify:

```
• computational cell size/resolution
```

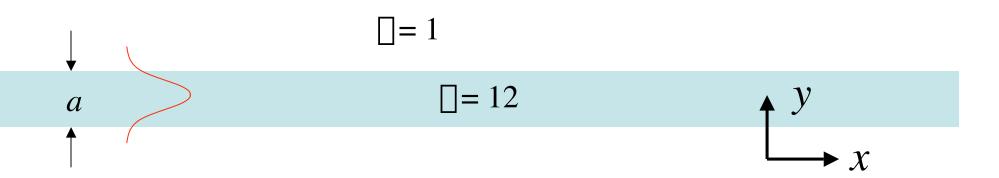
```
• geometry, i.e. \boxed{y}

• what k values

(set! k-points

(interpolate 10 (list (vector3 0 0 0) (vector3 2 0 0))))
```

• how many modes (n = 1, 2, ... ?)



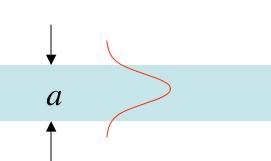
#### Need to specify:

- computational cell size/resolution
- geometry, i.e.  $\square(y)$
- what *k* values
- how many modes (n = 1, 2, ...?) (set! num-bands 5)

```
...Then run:
```

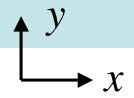
or only TM polarization: (run-tm)

or only TM, even modes: (run-tm-yeven)

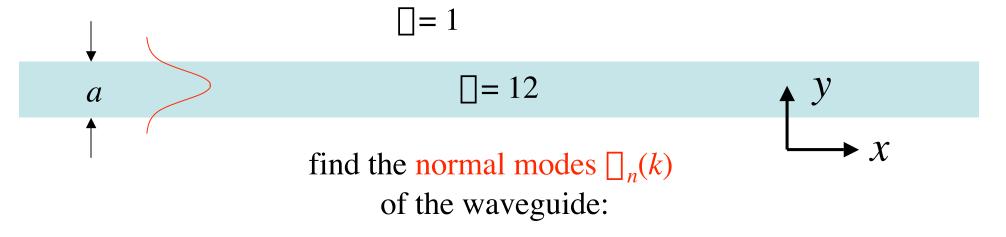


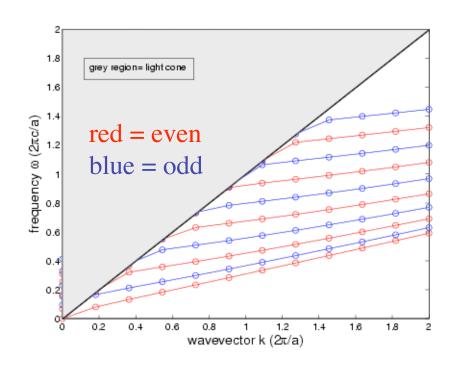
 $\square = 1$ 

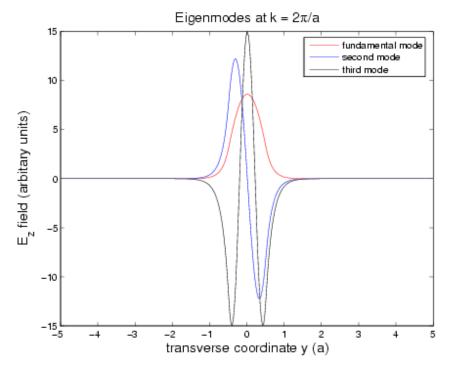
 $\square = 12$ 



### Simple Example (MPB) Results







#### Do try this at home

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