Computational EM Overview

Steven G. Johnson MIT course 18.369/8.315 Spring 2020

first, some perspective...

1 Analytical solutions

vacuum, single/double interfaces various electrostatic problems, ...





Lord Rayleigh

scattering from small particles, periodic multilayers (Bragg mirrors), ...

> ... & other problems with very high symmetry and/or separability and/or small parameters

Analytical solutions

Semi-analytical solutions: series expansions



Gustav Mie (1908)

e.g. Mie scattering of light by a sphere

Also called *spectral methods*:

Expand solution in rapidly converging Fourier-like basis

• spectral integral-equation methods:

exactly solve homogeneous regions (Green's func.),
& match boundary conditions via spectral basis
(e.g. Fourier series, spherical harmonics)

• spectral PDE methods:

spectral basis for unknowns in inhomogeous space(e.g. Fourier series, Chebyshev polynomials, ...)& plug into PDE and solve for coefficients

Analytical solutions

Semi-analytical solutions & spectral methods



Expand solution in *rapidly converging Fourier-like basis* e.g. Mie scattering of light by a sphere

Strength: can converge *exponentially fast*

- fast enough for hand calculation
- analytical insights, asymptotics, ...

Gustav Mie (1908) Limitation: fast ("spectral") convergence requires basis to be redesigned for each geometry (to account for any discontinuities/singularities ... complicated for complex geometries!)

(Or: brute-force Fourier series, polynomial convergence)

- 1) Analytical solutions
- 2) Semi-analytical solutions & spectral methods
- 3) Brute force: generic grid/mesh (or generic spectral)
- PDEs: discretize space into grid/mesh
 simple (low-degree polynomial) approximations in each pixel/element
 - ←finite differences (or Fourier series)
 - & finite elements \rightarrow



 boundary elements mesh surface unknowns coupled by Green's functions



lose orders of magnitude in performance … *but* re-usable code € computer time << €€€ programmer time

Computational EM: Three Axes of Comparison

- What *problem* is solved?
- eigenproblems: harmonic modes ~ $e^{-i\omega t}$ (**J** = 0)
- frequency-domain response: **E**, **H** from $J(x)e^{-i\omega t}$
- time-domain response: **E**, **H** from $J(\mathbf{x}, t)$
- PDE or integral equation?
- finite differences (FD)
- finite elements (FEM) / boundary elements (BEM)
- spectral / Fourier
- wns $-\ldots$
- What *solution method*?
- dense linear solvers (LAPACK)
- sparse-direct methods
- iterative methods

• What *discretization*? infinitely many unknowns

 \Rightarrow finitely many unknowns

A few lessons of history

- All approaches still in widespread use
 - brute force methods in 90%+ of papers, typically the first resort to see what happens in a new geometry
 - geometry-specific spectral methods still popular, especially when particular geometry of special interest
 - analytical techniques used less to solve new geometries than to prove theorems, treat small perturbations, etc.
- No single numerical method has "won" in general
 - each has strengths and weaknesses, e.g. tradeoff between simplicity/generalizability and performance/scalability
 - very mature/standardized problems (e.g. capacitance extraction) use increasingly sophisticated methods (e.g. BEM), research fields (e.g. nanophotonics) tend to use simpler methods that are easier to modify (e.g. FDTD)

Computing & Interpreting Band Structures & Dispersion Relations

> Steven G. Johnson MIT Applied Mathematics

Understanding Photonic Devices

[Xu & Lipson, 2005] [Notomi *et al.* (2005).] $10\mu m$ 420 nm [Mangan, *et al.*, OFC 2004 PDP24]

Model the whole thing at once? Too hard to understand & design.

Break it up into pieces first: periodic regions, waveguides, cavities

Building Blocks: "Eigenfunctions"

• Want to know what solutions exist in different regions and how they can interact: look for time-harmonic modes ~ $e^{-i\omega t}$

$$\vec{\nabla} \times \vec{E} = -\mu \frac{1}{\partial t} \frac{\partial}{\partial t} \vec{H} \rightarrow i\omega \vec{H}$$
First task:
get rid of this mess

$$\vec{\nabla} \times \vec{H} = \varepsilon \frac{\partial}{\partial t} \vec{E} + \vec{J} \stackrel{0}{\rightarrow} -i\omega \varepsilon \vec{E}$$



Building Blocks: Periodic Media



common thread:

translational symmetry

discrete periodicity: photonic crystals 1-D 2-D 3-D







periodic in three directions

periodic in one direction

periodic in two directions

Periodic Hermitian Eigenproblems

[G. Floquet, "Sur les équations différentielles linéaries à coefficients périodiques," Ann. École Norm. Sup. 12, 47–88 (1883).
 [F. Bloch, "Über die quantenmechanik der electronen in kristallgittern," Z. Physik 52, 555–600 (1928).

if eigen-operator is periodic, then Bloch-Floquet solutions:

can choose:
$$\vec{H}(\vec{x},t) = e^{i(\vec{k}\cdot\vec{x}-\omega t)}\vec{H}_{\vec{k}}(\vec{x})$$

planewave periodic "envelope"

Corollary 1: **k** is conserved, *i.e.* no scattering of Bloch wave Corollary 2: $\vec{H}_{\vec{k}}$ given by finite unit cell, $\circ \circ \circ$ so ω are discrete $\omega_n(\mathbf{k})$

Electronic and Photonic Crystals atoms in diamond structure | dielectric spheres, diamond lattice



Periodic Medium

Bloch waves: Band Diagram

strongly interacting fermions





wavevector weakly-interacting bosons ... many design degrees of freedom

Solving the Maxwell Eigenproblem

Finite cell \rightarrow *discrete* eigenvalues ω_n

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



$$(\nabla + i\mathbf{k}) \times \frac{1}{\varepsilon} (\nabla + i\mathbf{k}) \times \mathbf{H}_n = \frac{\omega_n^2}{c^2} \mathbf{H}_n$$

constraint: $(\nabla + i\mathbf{k}) \cdot \mathbf{H}_n = 0$

where field =
$$\mathbf{H}_{n}(\mathbf{x}) e^{i(\mathbf{k}\cdot\mathbf{x} - \omega t)}$$

1

Limit range of **k**: irreducible Brillouin zone

2 Limit degrees of freedom: expand **H** in finite basis



Solving the Maxwell Eigenproblem: 1

- **1** Limit range of \mathbf{k} : irreducible Brillouin zone
- -Bloch's theorem: solutions are periodic in **k** -Bloch's theorem: solutions are periodic in **k** -Bloch's theorem: solutions are periodic in **k** $\downarrow 2\pi$ $\downarrow 2\pi$ $\downarrow a$ $\downarrow 4y$ $\downarrow 4x$ \downarrow
 - 2
 -) Limit degrees of freedom: expand **H** in finite basis
 - 3
- Efficiently solve eigenproblem: iterative methods

Solving the Maxwell Eigenproblem: 2a Limit range of **k**: irreducible Brillouin zone Limit degrees of freedom: expand **H** in finite basis (N)2 $|\mathbf{H}\rangle = \mathbf{H}(\mathbf{x}_t) = \sum_{m=1}^{N} h_m \mathbf{b}_m(\mathbf{x}_t) \text{ solve: } \hat{A} |\mathbf{H}\rangle = \boldsymbol{\omega}^2 |\mathbf{H}\rangle$ m=1 $Ah = \omega^2 Bh$ finite matrix problem: *inner product:* Galerkin method: $\langle \mathbf{f} | \mathbf{g} \rangle = \int \mathbf{f}^* \cdot \mathbf{g}$ $A_{ml} = \langle \mathbf{b}_{m} | \hat{A} | \mathbf{b}_{l} \rangle \quad B_{ml} = \langle \mathbf{b}_{m} | \mathbf{b}_{l} \rangle$ Efficiently solve eigenproblem: iterative methods 3

Solving the Maxwell Eigenproblem: 2b

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



Limit degrees of freedom: expand H in finite basis

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

Planewave (FFT) basis

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

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

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



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

Finite-element basis

constraint, boundary conditions:

Nédélec elements

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

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



) Efficiently solve eigenproblem: iterative methods

Solving the Maxwell Eigenproblem: 3a

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



Efficiently solve eigenproblem: iterative methods

$$Ah = \omega^2 Bh$$

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(Np²) time for p eigenvectors

(p smallest eigenvalues)

Solving the Maxwell Eigenproblem: 3b

- $\frac{1}{2}$ Limit range of **k**: irreducible Brillouin zone
- 2 Limit degrees of freedom: expand **H** in finite basis



Efficiently solve eigenproblem: iterative methods

$$Ah = \omega^2 Bh$$

Many iterative methods:

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

Solving the Maxwell Eigenproblem: 3c

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



Efficiently solve eigenproblem: iterative methods

$$Ah = \omega^2 Bh$$

Many iterative methods:

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

for Hermitian matrices, smallest eigenvalue ω_0 minimizes:

variational / min–max theorem

$$\omega_0^2 = \min_h \frac{h^* A h}{h^* B h}$$

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



The Iteration Scheme is *Important* (minimizing function of 10⁴–10⁸+ variables!)

$$\omega_0^2 = \min_h \frac{h^* A h}{h^* B h} = f(h)$$

Steepest-descent: minimize $(h + \alpha \nabla f)$ over α ... repeat

Conjugate-gradient: minimize $(h + \alpha d)$ - *d* is ∇f + (stuff): *conjugate* to previous search dirs

Preconditioned steepest descent: minimize $(h + \alpha d)$ - $d = (approximate A^{-1}) \nabla f \sim Newton's method$

Preconditioned conjugate-gradient: minimize $(h + \alpha d)$ - *d* is (approximate A⁻¹) [∇f + (stuff)]

The Iteration Scheme is *Important* (minimizing function of ~40,000 variables)





The ε-averaging is Important



correct averaging changes *order* of convergence from Δx to Δx^2

reason in a nutshell:

 averaging
 smoothing ε
 changing structure
 must pick smoothing
 with zero 1st-order
 perturbation

[Farjadpour et al. (2006)]

Closely related to anisotropic metamaterial, e.g. multilayer film in large- λ limit

 $\lambda >> a$



key to anisotropy is differing continuity conditions on **E**:

 $\int E_{\parallel} \text{ continuous } \Rightarrow \epsilon_{\parallel} = \langle \epsilon \rangle$

 $D_{\perp} = \varepsilon E_{\perp}$ continuous $\Rightarrow \varepsilon_{\perp} = \langle \varepsilon^{-1} \rangle^{-1}$

Intentional "defects" are good

microcavities



waveguides ("wires")



Intentional "defects" in 2d

(Same computation, with supercell = many primitive cells)

microcavities a

(boundary conditions ~ irrelevant for exponentially localized modes) waveguides



Air-waveguide Band Diagram



any state in the gap cannot couple to bulk crystal -> localized

to be continued...





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Further reading:

Photonic Crystals book: <u>http://jdj.mit.edu/book</u>

Bloch-mode eigensolver: <u>http://jdj.mit.edu/mpb</u>

Computational Nanophotonics: Cavities and Resonant Devices

> Steven G. Johnson MIT Applied Mathematics

Resonance

an oscillating mode trapped for a long time in some volume (of light, sound, ...) lifetime $\tau >> 2\pi/\omega_0$ frequency ω_0 quality factor $Q = \omega_0 \tau/2$ energy in cavity ~ $e^{-\omega_0 t/Q}$ volume V



Resonance = Pole in Green's Function

an oscillating mode trapped for a long time in some volume (of light, sound, ...) lifetime $\tau >> 2\pi/\omega_0$ modal frequency ω_0 quality factor $Q = \omega_0 \tau/2$ volume V energy in cavity ~ $e^{-\omega_0 t/Q}$ ~ volume where residue is large Im ω near ω_0 , Green's function is dominated by contribution of the pole \sim a "resonant mode" profile **→** Re ω Х simple pole response to a narrowband pulse at $\omega_0 - i/\tau$ ~ exponential decay in time [causality/passivity: (in vicinity of the cavity) poles only for Im $\omega \le 0$]

Green's functions, briefly

Green's function = field(s) at x from dipole at y at a frequency ω

 $(\nabla \times \mu^{-1} \nabla \times - \omega^{2} \varepsilon) \mathbf{E}^{(j)}(\mathbf{x}) = i\omega \delta(\mathbf{x} - \mathbf{y}) \times (\text{unit vector in } j)$ $\Rightarrow \text{ electric "dyadic" Green's function } \mathbf{G}_{\omega}(\mathbf{x}, \mathbf{y}) = \begin{bmatrix} \mathbf{E}^{(1)} & \mathbf{E}^{(2)} & \mathbf{E}^{(3)} \end{bmatrix}$

> ... any electric current $\mathbf{J}(\mathbf{x})e^{-i\omega t}$ then gives the "convolution" $\mathbf{E}(\mathbf{x}) = \mathbf{G}_{\omega} * \mathbf{J} = \int \mathbf{G}_{\omega}(\mathbf{x}, \mathbf{y})\mathbf{J}(\mathbf{y})d^{3}\mathbf{y}$

At eigenvalue/resonance frequency $\omega_0 (\nabla \times \mu^{-1} \nabla \times \mathbf{E}_0 = \omega_0^2 \varepsilon \mathbf{E}_0)$, the operator $(\nabla \times \mu^{-1} \nabla \times -\omega_0^2 \varepsilon)$ becomes singular. G_{ω} blows up = "pole" at ω_0^2

> Similarly, 6×6 Green's function $\Gamma_{\omega}(\mathbf{x}, \mathbf{y})$ gives $\begin{bmatrix} \mathbf{E} \\ \mathbf{H} \end{bmatrix} = \psi$ fields from 6-component currents $\xi = \begin{bmatrix} \mathbf{J} \\ \mathbf{K} \end{bmatrix}$ at via $\psi = \Gamma_{\omega} * \xi$.

Microcavity Blues



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

- Best methods compute lowest-ω eigenvals, but N^d supercells have N^d 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 (1) source

 $\Delta \omega$

decay rate in time gives loss

FDTD: finite difference time domain

Finite-difference-time-domain (FDTD) is a method to model Maxwell's equations on a **discrete time** & **space grid** using finite centered differences

$$\nabla \times \mathbf{E} = -\frac{\partial \mathbf{B}}{\partial t}$$
 $\nabla \times \mathbf{H} = \frac{\partial \mathbf{D}}{\partial t} + \mathbf{J}$

 $\mathbf{D} = \varepsilon \mathbf{E}$ $\mathbf{B} = \mu \mathbf{H}$



K.S. Yee 1966

A. Taflove & S.C. Hagness 2005

FDTD: Yee leapfrog algorithm

2d example:

 at time t: Update D fields everywhere using spatial derivatives of H, then find E=ε⁻¹D

$$\mathbf{E}_{\mathbf{X}} \coloneqq \frac{\Delta t}{\varepsilon \Delta y} \left(\mathbf{H}_{\mathbf{Z}}^{j+0.5} - \mathbf{H}_{\mathbf{Z}}^{j-0.5} \right)$$

$$\mathbf{E}_{\mathbf{y}} \coloneqq \frac{\Delta t}{\varepsilon \Delta \mathbf{X}} \left(\mathbf{H}_{\mathbf{Z}}^{i+0.5} - \mathbf{H}_{\mathbf{Z}}^{i-0.5} \right)$$

2) at time t+0.5: Update H fields everywhere using spatial derivatives of E

$$\mathbf{H}_{z} \coloneqq \frac{\Delta t}{\mu} \left(\underbrace{\mathbf{E}_{\mathbf{X}}^{j+1} - \mathbf{E}_{\mathbf{X}}^{j}}_{\Delta y} + \underbrace{\mathbf{E}_{\mathbf{y}}^{i} - \mathbf{E}_{\mathbf{y}}^{i+1}}_{\Delta \mathbf{x}} \right)$$

CFL/Von Neumann stability: $c\Delta t < 1 / \sqrt{\Delta x^{-2} + \Delta y^{-2}}$



Free software: MEEP

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

- FDTD Maxwell solver: 1d/2d/3d/cylindrical
- Parallel, scriptable, integrated optimization, signal processing
- Arbitrary geometries, anisotropy, dispersion, nonlinearity
- Bloch-periodic boundaries, symmetry boundary conditions, + PML absorbing boundary layers...



Absorbing boundaries?

Finite-difference/finite-element volume discretizations need to artificially truncate space for a computer simulation.



In a wave equation, a hard-wall truncation gives reflection artifacts.

An old goal: "absorbing boundary condition" (ABC) that absorbs outgoing waves.

Problem: good ABCs are hard to find in > 1d.

Perfectly Matched Layers (PMLs)

Bérenger, 1994: design an *artificial* absorbing layer that is analytically reflectionless



Works remarkably well.

Now ubiquitous in FD/FEM wave-equation solvers.

Several derivations, cleanest & most general via "complex coordinate stretching" [Chew & Weedon (1994)]

Perfectly Matched Layers (PMLs)

Bérenger, 1994: design an *artificial* absorbing layer that is analytically reflectionless



Even works in inhomogeneous media (e.g. waveguides).

PML Starting point: propagating wave

• Say we want to absorb wave traveling in +x direction in an x-invariant medium at a frequency $\omega > 0$.

fields ~
$$f(y,z)e^{i(kx-\omega t)}$$



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(usually, k > 0)
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[rare "backward-wave" cases defeat PML (Loh, 2009)]

(only x in wave equation is via $\partial / \partial x$ terms.)

PML step 1: Analytically continue

Electromagnetic fields & equations are *analytic* in x, so we can evaluate at complex x & still solve same equations



PML step 2: Coordinate transformation

Weird to solve equations for complex coordinates \tilde{x} , so do coordinate transformation back to real x.

$$\tilde{x}(x) = x + \int_{-\infty}^{x} \frac{i\sigma(x')}{\omega} dx'$$

(allow *x*-dependent PML strength σ)

$$\frac{\partial}{\partial x} \xrightarrow{1} \frac{\partial}{\partial \tilde{x}} \xrightarrow{2} \left[\frac{1}{1 + \frac{i\sigma(x)}{\omega}} \right] \frac{\partial}{\partial x}$$

fields ~ $f(y,z)e^{i(kx-\omega t)} \rightarrow f(y,z)e^{i(kx-\omega t)}$ nondispersive materials: k/ω ~ constant

so decay rate independent of ω

(at a given incidence angle)

PML Step 3: Effective materials

In Maxwell's equations, $\nabla \times \mathbf{E} = i\omega\mu\mathbf{H}$, $\nabla \times \mathbf{H} = -i\omega\varepsilon\mathbf{E} + \mathbf{J}$, coordinate transformations are *equivalent* to transformed *materials* (Ward & Pendry, 1996: "transformational optics")

$$\{\varepsilon,\mu\} \rightarrow \frac{J\{\varepsilon,\mu\}J^T}{\det J}$$

x PML Jacobian

 $J = \begin{pmatrix} (1 + i\sigma / \omega)^{-1} & & \\ & 1 & \\ & & 1 \end{pmatrix}$

for isotropic starting materials: effective

$$\{\varepsilon,\mu\} \rightarrow \{\varepsilon,\mu\} \begin{pmatrix} (1+i\sigma/\omega)^{-1} & 0 \\ 1+i\sigma/\omega & 0 \\ 1+i\sigma/\omega & 0 \end{pmatrix}$$



PML = effective anisotropic "absorbing" ε , μ

Photonic-crystal PML?

FDTD (Meep) simulation:



ε not even continuous in x direction, much less analytic!

Analytic continuation of Maxwell's equations is hopeless — no reason to think that PML technique should work

Photonic-crystal PMLs: Magic?

[Koshiba, Tsuji, & Sasaki (2001)]

[Kosmidou et al (2003)]



... & several other authors ...
Low reflections claimed

is PML working?

Something suspicious:

very thick absorbers.

Failure of Photonic-crystal "pseudo-PML"

[Oskooi et al, Optics Express 16, 11376 (2008)]



Redemption of the pseudo-PML: *slow ("adiabatic") absorption turn-on*

[Oskooi et al, Optics Express 16, 11376 (2008)]



Any absorber, turned on gradually enough, will have reflections $\rightarrow 0$!

PML (when it works) just helps coefficient.

What about DtN / RCWA / Blochmode-expansion / SIE methods?

 useful, nice methods that can impose outgoing boundary conditions exactly, once the Green's function / Bloch modes computed

challenge problem for any method: periodic 3d dielectric waveguide bend in air (note: both guided and radiating modes!)



... DtN / Green's function / Bloch modes (incl. radiation!) expensive

Computational Nanophotonics: Sources & Integral Equations

> Steven G. Johnson MIT Applied Mathematics

How can we excite a desired incident wave?



Want some current source to excite incident waveguide mode, planewave, etc...

 also called transparent source since waves do not scatter from it (thanks to linearity)

— vs. hard source = Dirichlet field condition

Equivalent currents ("total-field/scattered-field" approach)

[review article: arXiv:1301.5366]



equivalent currents \mathbf{c} $\mathbf{\Omega}$ \mathbf{f}^+ $\mathbf{f}=0$



known incident fields

 $\mathbf{f}^{+} = \left(\begin{array}{c} \mathbf{E} \\ \mathbf{H} \end{array}\right)$

in ambient medium

(possibly inhomogeneous, e.g. waveguide or photonic crystal) want to construct surface currents

$$\mathbf{c} = \left(\begin{array}{c} \mathbf{J} \\ \mathbf{K} \end{array}\right)$$

giving same f^+ in Ω

do simulations in finite domain + inhomogeneities / interactions

= scattered field **f**-

The *Principle of Equivalence* in classical EM

(or Stratton–Chu equivalence principle)
 (formalizes Huygens' Principle)
 (or total-field/scattered-field, TFSF)
 (near-to-far-field transformation)

(close connection to Schur complement [Kuchment])

[see e.g. Harrington, *Time-Harmonic Electromagnetic Fields*] [review article: arXiv:1301.5366]

starting point: solution in all space



solve (source-free) Maxwell PDE (in frequency domain):

$$\begin{pmatrix} \nabla \times \\ -\nabla \times \end{pmatrix} \mathbf{f}^{+} = -i\omega\chi\mathbf{f}^{+}$$

constructing solution in Ω



construct **c** so that **f** is a new solution:

$$\begin{pmatrix} \nabla \times \\ -\nabla \times \end{pmatrix} \mathbf{f} = -i\omega\chi\mathbf{f} + \delta(\partial\Omega) \begin{pmatrix} -n \times \mathbf{H}^+ \\ n \times \mathbf{E}^+ \end{pmatrix}$$
 "electric" current
= $-i\omega\chi\mathbf{f} + \mathbf{c}$

Exciting a waveguide mode in FDTD

– compute mode in MPB, then use as source in MEEP



[review article: arXiv:1301.5366]

Problems with equivalent sources

(if not solved, undesired excitation of other waves) [review article: arXiv:1301.5366]

- Discretization mismatch: at finite resolution, solutions from one technique (MPB) don't exactly match discrete modes in another technique (Meep) — leads to small imperfections — solvable by using the same discretization to find modes
- Dispersion: mode profile varies with ω , so injecting a pulse p(t) requires a convolution with $\hat{\mathbf{c}}(\mathbf{x},\omega)_{\text{Fourier}} \mathbf{c}(\mathbf{x},t)$

 $\operatorname{currents}(\mathbf{x},t) = p(t) * \mathbf{c}(\mathbf{x},t) \approx p(t) \, \hat{\mathbf{c}}(\mathbf{x},\omega)$ *narrow-bandwidth*

- convolutions expensive, can be approximated by finite-time (FIR/IIR) calculations using DSP techniques
- specialized methods are known for planewave sources (have numerical dispersion!)



Shortcut: Subtract two simulations

example: 90° bend of single-mode dielectric waveguide





trick #1: incident & scattered fields are Bloch-periodic/quasiperiodic

trick #2: *e^{ik_xx* current source produces planewave}

Reflection spectra example for periodic media

(Fano resonance lineshapes)



in (ω, θ) plot



+ normalization run

Fun possibilities in FDTD: moving sources [= just some currents J(x,t)]



Doppler shift from moving oscillating dipole

$v = 1.05 c/n (0.35 pixels/\Delta t)$

Cerenkov radiation from moving point charge in dielectric medium

Cerenkov radiation



 $v = 1.05 c/n (0.35 pixels/\Delta t)$

excites radiating mode $\omega(k_x, k_y)$ if $v = \omega(k_x, k_y)/k_x$ = phase velocity in x direction $\ge c/n$ in index-n medium

Cerenkov radiation in photonic crystal



excites radiating mode $\omega(k_x,k_y)$ if $v = \omega(k_x,k_y)/(k_x + 2\pi m/a)$ for any integer m

⇒ no minimum v
[Smith–Purcell effect]



very different radiation patterns & directions depending on *v*, due to interactions with 2d PhC dispersion curves

[Luo, Ibanescu, Johnson,& Joannopoulos (Science, 2002)]

Surface-integral equations (SIEs) and boundary-element methods (BEMs)

[see e.g. Harrington, Time-Harmonic Electromagnetic Fields]

Harrington, "Boundary integral formulations for homogeneous material bodies," J. Electromagnetic Waves Appl. 3, 1–15 (1989)

Chew et al., Fast and Efficient Algorithms in Computational Electromagnetics (2001)].

Exploiting partial knowledge of Green's functions

a typical scattering problem:



suppose that we know Green's functions in infinite medium 0 or medium 1

- known analytically for homogeneous media
- computable by *much smaller* calculation in periodic medium

Can exploit this to derive integral equation for surface unknowns only.

The *Principle of Equivalence* in classical EM

[see e.g. Harrington, Time-Harmonic Electromagnetic Fields]



Maxwell PDE:

$$\begin{pmatrix} \nabla \times \\ -\nabla \times \end{pmatrix} \mathbf{f} = -i\omega \boldsymbol{\chi}^{(0,1)} \mathbf{f}$$

... we want to partition *into* separate *medium* 0/1 *problems* & enforce continuity...

Constructing a medium-0 solution



$$\begin{pmatrix} \nabla \times \\ -\nabla \times \end{pmatrix} \mathbf{f} = -i\omega\chi\mathbf{f} + \delta(\partial\Omega) \begin{pmatrix} -n \times \mathbf{H}^0 \\ n \times \mathbf{E}^0 \end{pmatrix}$$
 "electric" current
= $-i\omega\chi\mathbf{f} + \mathbf{c}$

The Principle of Equivalence I



[e.g. Harrington, Time-Harmonic Electromagnetic Fields]
The Principle of Equivalence II



Surface-Integral Equations (SIE)



Discretizing the Maxwell SIE



Discretizing the Maxwell SIE





... + straightforward generalizations to more objects, nested objects, etcetera

SIE basis choices

Can use *any* basis for c = any basis of surface functions
 ... basis is *not* incoming/outgoing waves
 & need *not* satisfy *any wave equation*

- Spectral bases: spherical harmonics, Fourier series, nice for high symmetry
 - ~ uniform spatial resolution
- Boundary Element Methods (BEM):

localized basis functions defined on irregular mesh





"RWG" basis (1982):

vector-valued \mathbf{b}_n defined on *pairs* of adjacent triangles via degree-1 polynomials

BEM strengths

especially small surface areas in a large (many- λ) volume, e.g.:

surface plasmons (metals): extremely sub- λ fields



[Johannes Feist, Harvard]





Graphene

~ delta-function surface conductivity = jump discontinuity $(\sim E)$ in **H** field

The bad news of BEM

- Not well-suited for nonlinear, time-varying, or non-piecewise-constant media
- BEM system matrix $M_{mn} = \left\langle \mathbf{b}_{m} \middle| \left(\Gamma^{0} + \Gamma^{1} \right) * \mathbf{b}_{n} \right\rangle = G_{mn}^{0} + G_{mn}^{1}$

- *singular* integrals for overlapping \mathbf{b}_m , \mathbf{b}_n ... special numerical integration techniques

— *M* is *not sparse*, but:

often small enough for dense solvers ($\leq 10^4 \times 10^4$) + "fast solvers:" approximate sparse factorizations (fast multipole method, etc.)

lots of work every time you change Γ

 (e.g. 3d vs. 2d, periodic boundaries, anisotropic, ...)
 but independent of geometry

The good news of BEM: You don't have to write it yourself



Free software developed by Dr. Homer Reid (collaboration with Prof. Jacob White @ MIT)



SCUFF-EM

[https://github.com/HomerReid/scuff-em]

Surface-CUrrent / Field Formulation of Electro-Magnetism

SCUFF-EM is a free, open-source software implementation of the boundary-element method of electromagnetic scattering.

SCUFF-EM supports a wide range of geometries, including compact scatterers, infinitely extended scatterers, and multi-material junctions.

The SCUFF-EM suite includes 8 standalone application codes for specialized problems in EM scattering, fluctuation physics, and RF engineering.

The SCUFF-EM suite also includes a core library with C++ and PYTHON APIs for designing homemade applications.

https://github.com/HomerReid/scuff-em

SCUFF usage outline

The steps involved in solving any BEM scattering problem:

1. Mesh object surfaces into triangles.

Not done by SCUFF-EM; high-quality free meshing packages exist (e.g. GMSH).

- 2. Assemble the BEM matrix M and RHS vector v. SCUFF-EM does this.
- 3. Solve the linear system Mk = v for the surface currents k. SCUFF-EM USES LAPACK for this.
- 4. Post-process to compute scattered fields $\{E, H\}^{\text{scat}}$ from k. SCUFF-EM does this.
- Innovations unique to SCUFF-EM:
 - Bypass step 4: Compute scattered/absorbed power, force, and torque directly from k
 - Bypass steps 3 and 4: Compute Casimir forces and heat transfer directly from M

Geometries in SCUFF

A gold sphere and a displaced and rotated SiO2 tetrahedron:



The .scuffgeo file:

OBJECT TheSphere MESHFILE Sphere.msh MATERIAL Gold ENDOBJECT

OBJECT ThePyramid MESHFILE Pyramid.msh MATERIAL SiO2 DISPLACED 0 0 -1 ROTATED 45 ABOUT 0 1 0 ENDOBJECT

⇒ Handle displacements and rotations without re-meshing.

Geometries in SCUFF



(discretization of SIE at junctions of 3+ materials is a bit tricky)

Periodic geometries in SCUFF



(implementing periodicity is nontrivial: changes Green's function! SCUFF: periodic $\Gamma = \Sigma$ (nearest neighbors) + Ewald summation)

Using SIE/BEM solutions

Solving the SIE gives the surface currents \mathbf{c} , and from these (via $\Gamma^*\mathbf{c}$) one can obtain any desired fields, but...

It is much more efficient to compute as much as possible directly from $c (\sim n \times \text{surface fields})$. Examples:

- Scattering matrices (e.g. spherical-harmonic waves in → out): obtain directly from multipole moments of "currents"
- Any bilinear function of the surface fields can be computed directly from bilinear functions of **c**:
 - scattered/absorbed power, force, torque, ...

https://arxiv.org/abs/1307.2966

• Net effects of quantum/thermal fluctuations in matter can be computed from norm/det/trace of M or M⁻¹:

- thermal radiation, Casimir (van der Waals) forces, ...

Resonant modes (and eigenvalues)

• BEM scattering problems are of the form $M(\omega)x = s$. Resonances (and eigenvalues) are ω where this system is singular, i.e. the **nonlinear eigenproblem**

 $\det M(\omega) = 0$

For passive (\Rightarrow causal) systems, solutions can only occur for Im $\omega \le 0$.

• Various algorithms exist, including an intriguing algorithm using contour integrals of $M(\omega)$ [Beyn (2012)].

to be continued...

Further reading:

Free FDTD software: <u>http://jdj.mit.edu/meep</u> Free BEM software:

http://homerreid.ath.cx/scuff-EM/

Review on wave sources:

arXiv:1301.5366 [in Taflove, Oskooi, & Johnson, eds., Advances in FDTD Comput. EM (2013)] Computational Nanophotonics: Optimization and "Inverse Design"

> Steven G. Johnson MIT Applied Mathematics

Many, many papers that parameterize by a *few* degrees of freedom and optimize...

Today, focus is on *large-scale optimization*, also called *inverse design*: so many degrees of freedom (10²–10⁶) that computer is "discovering" new designs.

Outline

- Brief overview/examples of large-scale optimization work in photonics
- Overview of optimization terminology, problem types, and techniques.
- Some more detailed photonics examples.

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Optical design = optimization

traditional approach: intuition + "tweaking" few parameters









[Noda et. al. 2003] [Brongersma et. al. 2010] [Yu et. al. 2010] [Zhou et. al. 2010]

gradient-based ("adjoint") optimization (>10⁵ params, 3D)



[Sigmund et. al. *Las. Phot. Rev.* 5, 308 (2011)]

[X. Liang & SG Johnson *Opt. Exp.* 21, 30812 (2013)]

"black-box" optimization (typically << 100 params)



[Villegas et. al. 2004] [Hakansson et. al. 2005]

Large-scale optimization in photonics: "Every pixel" is a degree of freedom

solar-cell backreflector optimization

bend optimization



Sigmund et al., Opt. Express **12**, 1996 (2004)





710 nm

Ganapati et al. IEEE Jour. of Photovolt. 4, 175 (2014)



OE 12, 5916 (2004)



Topology optimization



Given two (or more) materials A and B, determine what arrangement — including what topology optimizes some objective/constraints.

Electromagnetism: Materials (mostly) described by permittivity (dielectric constant) ϵ (susceptibility $\chi = \epsilon - 1$)

Discretizing Topology Optimization

for computer, need finite-dimensional, differentiable parameters

some computational grid



(+ filtering methods to constrain minimum feature sizes and "binary-ize" result) Level-set method: value of "level-set" function $\phi(\mathbf{x})$ varies continuously at each pixel \Rightarrow material *A* or *B* if $\phi > 0$ or < 0

...*or* ...

"Density-based topology optim." Continuous relaxation: material varies in [*A*,*B*] at each pixel

e.g. in electromagnetism, let ε at each pixel vary in [*A*,*B*].

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A general optimization problem

 $\min_{x\in\mathbb{R}^n}f_0(x)$

subject to *m* constraints

 $f_i(x) \le 0$

i = 1, 2, ..., m

x is a *feasible point* if it satisfies all the constraints *feasible region* = set of all feasible *x*

minimize an objective function f_0 with respect to *n* design parameters *x* (also called *decision parameters, optimization variables*, etc.)

> - note that *maximizing* g(x)corresponds to $f_0(x) = -g(x)$

note that an *equality constraint* h(x) = 0yields two inequality constraints $f_i(x) = h(x)$ and $f_{i+1}(x) = -h(x)$ (although, in practical algorithms, equality constraints typically require special handling)

Important considerations

• Global versus local optimization

photonics: mostly local optima in

- *Convex* vs. non-convex optimization non-convex problems
- Unconstrained or box-constrained optimization, and other special-case constraints
- Special classes of functions (linear, etc.)
- Differentiable vs. non-differentiable functions
- Gradient-based vs. derivative-free algorithms
- Zillions of different algorithms, usually restricted to various special cases, each with strengths/weaknesses

Relaxations of Integer Programming

If *x* is integer-valued rather than real-valued (e.g. $x \in \{0,1\}^n$), the resulting *integer programming* or *combinatorial optimization* problem becomes *much harder* in general (often NP-complete).

However, useful results can often be obtained by a *continuous relaxation* of the problem — e.g., going from $x \in \{0,1\}^n$ to $x \in [0,1]^n$

... at the very least, this gives an lower bound on the optimum f_0 ... and penalty methods (e.g. SIMP) can be used to gradually eliminate intermediate x values.

Leads to "density based" topology optimization, many methods to impose feature-size constraints etc.

Derivatives are essential

$$\min_{x \in \mathbb{R}^n} f_0(x)$$

subject to *m* constraints
$$f_i(x) \leq 0$$

 $i = 1, 2, ..., m$

For $n \ge 1000$'s of parameters, impractical unless you have

$$\nabla_x f_i(x)$$

$$i=0,1,2,\ldots,m$$

computed "analytically" (not by finite differences).

minimize an objective function f_0 with respect to *n* design parameters *x* (also called *decision parameters, optimization variables*, etc.) Impossible to explore/optimize a 10⁶-dimensional parameter space without derivatives.

(Gradient tells you which direction to go for improvement.)

(Only local optimization with this many parameters, but can still find very good designs, sometimes with provable guarantees.)

Amazing fact of adjoint methods: all 10⁶ derivatives with two simulations

physical intuition: Born approximation + reciprocity



"forward" solve

scattered field + perturbation ΔE = field of $J = \Delta \epsilon E_0$

perturbed pixel $\Delta \varepsilon$, *expensive: repeat for each pixel*?

Amazing fact of adjoint methods: all 10⁶ derivatives with two simulations

physical intuition: Born approximation + reciprocity

scattered field source at scattered + perturbation ΔE measurement point = field of $J = \Delta \epsilon E_0$ (reciprocity) perturbed pixel $\Delta \varepsilon$, solve one adjoint problem repeat for each pixel? ... get fields at **all** perturbed pixels Adjoint methods, in math $\cot \nabla f \sim \cot (x)$ evaluation [google "adjoint method" for reviews]

toy example: maximizing transmitted power from a source

Maxwell's equations discretized as: [real variables, e = real/imag parts] Quadratic objective: $f(x) = e^T Q e^{T} Q e^{T}$ (particular) [Q assumed symmetric]

$$M(x)^{\text{EM}} = S^{\text{source}}$$

Maxwell matrix (parameters x)

$$\frac{\partial f}{\partial x_i} = 2e^T Q \frac{\partial e}{\partial x_i} = -2e^T Q M^{-1} \frac{\partial M}{\partial x_i} e = 2a^T \frac{\partial M}{\partial x_i} e$$

adjoint problem: $M^T a = Qe$

= one extra solve with transposed (adjoint) M (Don't let the reciprocity intuition fool you.)

There is a general prescription that is independent of the physics — even for nonreciprocal, nonlinear, and time-varying problems.

(google "adjoint method notes")

(also known as "reverse mode" differentiation or, in machine learning, as "backpropagation")

Sometimes, non-obvious transformations are required to make the problem differentiable.
Designing photonic band gaps

periodic structures ("photonic crystals") have

Bloch-wave "quasiperiodic" solutions = periodic(x) × $e^{ikx-i\omega t}$



[Y.A. Vlasov et al., Nature 414, 289 (2001).]

wavevector k

In the gap, crystal is "optical insulator" that can trap light.

Maximizing photonic band gap over all periodic structures?

we want:
$$\max_{\varepsilon} \left(2 \frac{\left[\min_{\mathbf{k}} \omega_{n+1}(\mathbf{k}) \right] - \left[\max_{\mathbf{k}} \omega_{n}(\mathbf{k}) \right]}{\left[\min_{\mathbf{k}} \omega_{n+1}(\mathbf{k}) \right] + \left[\max_{\mathbf{k}} \omega_{n}(\mathbf{k}) \right]} \right)$$

frequently not differentiable

an equivalent problem ("epigraph" transformation for "minimax" problems):

...with (mostly?) differentiable nonlinear constraints:

$$\max_{\varepsilon, f_1, f_2} \left(2 \frac{f_2 - f_1}{f_2 + f_1} \right)$$

subject to:

 $f_1 \ge \boldsymbol{\omega}_n(\mathbf{k})$ $f_2 \le \boldsymbol{\omega}_{n+1}(\mathbf{k})$

for $\mathbf{k} \in \mathcal{K}$

Optimizing 1st TM and TE gaps for a triangular lattice with 6-fold symmetry (between bands 1 & 2)



48.3% TM gap ($\varepsilon = 12:1$)



51.4% TE gap (ε = 12:1)

30 iterations of optimizer

Optimizing 1st complete (TE+TM) 2d gap



21.1% gap ($\varepsilon = 12:1$)

+ some local minima



good news: only a handful of minima (in 10³-dimensional space!)

3d gap optimization [given symmetry group + which bands]



(e) FCC8 (no. 225)

~ $100 \times 100 \times 100 = 10^6$ degrees of freedom (ϵ in every "voxel")

Gap vs Index Contrast



Key questions occur *before* choosing optimization algorithm:

- How to parameterize the degrees of freedom — how much knowledge of solution to build in?
- Which objective function & constraints? — many choices for a given design goal,

... can make an enormous difference in the computational feasibility & the robustness of the result.

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3d Microcavity Design Problem

420 nm

[Song, (2005)]

000

0000

410 nm

00000000

radiation loss (finite \mathbf{Q}_{rad})

Want some 2d pattern that will confine light in 3d with maximal lifetime (" Q_{rad} ") and minimal modal volume ("V")

Many *ad hoc* designs, trading off Q_{rad} and V...

ring resonators

[Loncar, 2002]



("defects" in periodic structures)



[[]Akahane, 2003]

Resonances = complex $\omega \sim$ frequency - *i* loss



resonances = poles in scattering = poles in Green's function = singular Maxwell operator M(ω)

 $(\nabla \times \nabla \times -\omega^2 \varepsilon)\mathbf{E} = \mathbf{i}\omega\mathbf{J} = \mathbf{0}$ M(\omega) singular at resonance \omega

Rybin et. al (2017): arXiv:1706.02099

Optimize resonances?



Challenges:

Which eigenvalue?

Interior eigenvalue of big non-Hermitian...

Tracking eigenvalue (no discontin. jumps!)

Optimize resolvent instead



[X. Liang & S. G. Johnson, Optics Express (2013).]

Back to cavity optimization...

Typical figure of merit is "Purcell factor" Q/V[r(~ enhancement of light-matter coupling)ar

[review: arXiv:1301.5366]

= approximation for LDOS (local density of states) = power expended by dipole source Re $\psi^*M(\omega_0)^{-1}\psi$ for ψ = dipole

Naively, should we maximize Q/V or LDOS?

Trivial design problem: maximum Q/V = ∞
[for lossless materials,
e.g. perfect ring resonator of ∞ radius]



 $V \sim R$

Real design problem: maximize LDOS $Q_{rad} \sim exp(\# R)$ averaged over desired bandwidth $\omega_0 \pm \Gamma_0$ [Liang & Johnson (2013)]

LDOS: Local Density of States

[review: arXiv:1301.5366]

Maxwell eigenproblem:

Maxwell vector-Helmholtz:

 $\frac{1}{\varepsilon} \nabla \times \frac{1}{\mu} \nabla \times E \triangleq \Theta E = \omega^2 E \qquad E = i\omega (\Theta - \omega^2)^{-1} \varepsilon^{-1} J$ $\langle E, E' \rangle = \int E^* \cdot \varepsilon E'$

Power radiated by a current J (Poynting's theorem)

$$P = -\frac{1}{2} \operatorname{Re} \int \boldsymbol{E}^* \cdot \boldsymbol{J} \, d^3 \boldsymbol{x} = -\frac{1}{2} \operatorname{Re} \left\langle \boldsymbol{E}, \, \boldsymbol{\varepsilon}^{-1} \boldsymbol{J} \right\rangle$$

special case of a dipole source: LDOS

$$\boldsymbol{J}(\boldsymbol{x}) = \boldsymbol{e}_{\ell} \,\delta(\boldsymbol{x} - \boldsymbol{x}_0) \qquad \text{LDOS}_{\ell}(\boldsymbol{x}_0, \boldsymbol{\omega}) = \frac{4}{\pi} \,\varepsilon(\boldsymbol{x}_0) P_{\ell}(\boldsymbol{x}_0, \boldsymbol{\omega})$$

Why a "density of states"

[review: arXiv:1301.5366]

consider a finite domain (periodic/Dirichlet) + small absorption

$$\frac{1}{\varepsilon} \nabla \times \frac{1}{\mu} \nabla \times E \triangleq \Theta E = \omega^2 E$$
$$\langle E, E' \rangle = \int E \cdot \varepsilon E'$$

countable eigenfunctions $\mathbf{E}^{(n)}$ and frequencies $\omega^{(n)} - i\gamma^{(n)}$

$$E = i\omega(\Theta - \omega^2)^{-1} \varepsilon^{-1} J$$
$$P = -\frac{1}{2} \operatorname{Re} \langle E, \varepsilon^{-1} J \rangle$$
$$\varepsilon^{-1} J = \sum_{n} E^{(n)} \langle E^{(n)}, \varepsilon^{-1} J \rangle$$

 $loss \rightarrow 0$: a localized measure of spectral density

$$LDOS_{\ell}(\boldsymbol{x},\boldsymbol{\omega}) = \sum_{n} \delta(\boldsymbol{\omega} - \boldsymbol{\omega}^{(n)}) \varepsilon(\boldsymbol{x}) |E_{\ell}^{(n)}(\boldsymbol{x})|^{2}$$

DOS(\overline{\verline

Complex target ω_0 = Frequency average

• Passivity/causality: $M(\omega)^{-1}$ analytic for Im $\omega > 0$ $f(\omega) = \psi^* M(\omega_0)^{-1} \psi$



[X. Liang & S. G. Johnson, *Optics Express* (2013).] [Owen Miller et. al. *Phys. Rev. Lett.* 112, 123903 (2014)]

Complex $\omega = \omega$ average: Lots of uses

3d optimization of absorbing particles (frequency-averaged absorbed+scattered power)



[Owen Miller et. al. (2014)]

Modeling Casimir/van der Waals force



integrating fluctuations over all ω = much nicer integral over Im ω

[Rodriguez et al., *Nature Photonics* (2011)]

- General derivation of Wheeler–Chu limits via contour integration [Sohl, Gustafsson, Kristensson (2007)]
- Extension of "Miller" bounds to finite bandwidth [Shim (2019)]
- Proof that cloaking bandwidth scales ~ 1/diameter [Hashemi (2010)]

in-plane J

Jy

Maximizing LDOS for random in-plane J = max[LDOS(ω ,J_x)+LDOS(ω ,J_y)]/2





6 out of 10

Spontaneous symmetry breaking! "Picks" one polarization randomly

3d results: Photonic-crystal slab

[X. Liang & S. G. Johnson, Optics Express (2013).]



Optimize with $Q_0=10^4$

i.e. prefer $Q \ge 10^4$ but after that mainly minimize V

Next: 2d pattern in 3d slab

(including radiation loss via PML absorbing boundaries)





 $Q \sim 30,000, V \sim 0.06(\lambda/n)^3$

vs. hand-optimized: $Q \sim 100,000, V \sim 0.7(\lambda/n)^3$ $Q \sim 300,000, V \sim 0.2(\lambda/n)^3$ *and others...*

Manufacturability: Feature-size constraints

[Wang, Christiansen, Yu, Mørk, and Sigmund, Appl. Phys. Lett. 113, 241101 (2018)]

Various techniques to impose a minimum feature size, connectivity, and other manufacturing constraints in TO.





Hand-design SHG cavity [Bi et al. (2012)]





~80–90% efficiency (2d & 3d) for AlGaAs, 30mW power, at teleco wavelengths with 0.1% bandwidth

months of hand-tuning to find compatible resonance modes

SHG by "LDOS" optimization

[Lin, Liang, Lončar, Johnson, and Rodriguez, Optica, vol. 3, pp. 233–238 (2016).]

key idea:

source J_1 at ω_1 $\Rightarrow E_1$ $\Rightarrow J_2 \sim \chi^{(2)} E_1^2$ $\Rightarrow E_2$ where \Rightarrow power at ω_2

Maximizing SHG = maximizing composition of two scattering problems.

$$\max_{\tilde{e}_{\alpha}} \langle f(\tilde{e}_{\alpha}; \omega_{1}) \rangle = -\operatorname{Re} \left[\left\langle \int \mathbf{J}_{2}^{*} \cdot \mathbf{E}_{2} d\mathbf{r} \right\rangle \right],$$
$$\mathcal{M}_{1} \mathbf{E}_{1} = i\omega_{1} \mathbf{J}_{1},$$
$$\mathcal{M}_{2} \mathbf{E}_{2} = i\omega_{2} \mathbf{J}_{2}, \omega_{2} = 2\omega_{1}$$
$$\mathbf{J}_{1} = \delta(\mathbf{r}_{\alpha} - \mathbf{r}_{0}) \hat{\mathbf{e}}_{j}, \qquad j \in \{x, y, z\}$$
$$\mathbf{J}_{2} = \tilde{e}(\mathbf{r}_{\alpha}) E_{1j}^{2} \hat{\mathbf{e}}_{j},$$
$$\mathcal{M}_{l} = \nabla \times \frac{1}{\mu} \nabla \times -\varepsilon_{l}(\mathbf{r}_{\alpha}) \omega_{l}^{2}, \qquad l = 1, 2$$
$$\varepsilon_{l}(\mathbf{r}_{\alpha}) = \varepsilon_{m} + \tilde{\varepsilon}_{\alpha}(\varepsilon_{dl} - \varepsilon_{m}), \qquad \tilde{\varepsilon}_{\alpha} \in [0, 1],$$

SHG by "LDOS" optimization

[Lin, Liang, Lončar, Johnson, and Rodriguez, Optica, vol. 3, pp. 233–238 (2016).]



gave factor of 10 increase in mode-overlap figure of merit vs. best hand design

optimized VCSEL-like multilayer-film (~hundreds of degrees of freedom)

Topology optimization for nonlinear frequency conversion

(figs courtesy Z. Lin)



Surface-enhanced Raman Scattering [R. Christiansen, arXiv:1911.05002 (2019)]



enhance Raman both by focusing incident wave and by enhancing emission (Purcell effect) ... what structure is best?



60 × better than typical resonators [R. Christiansen, arXiv:1811.12936 (2019)]



"Metasurface" optical devices:



Large-area (often 100–1000 λ) nanopatterned surfaces designed to reflect/transmit desired waves — e.g. flat-lens focusing, beamforming,



(M.Khorasaninejad, W. T. Chen, R. C. Devlin, J.Oh, A. Y. Zhu ,F. Capasso, *Science*, June 3 2016)

M. Khorasaninejad et al.,. IEEE J. Sel. Top. Quantum Electron. 23, 4700216 (2017)



Capasso group (Harvar "Meta:" << λ pattern acts like effective surface "impedance". (Not really necessary.)



Complication: focus multiple incident λ and/or angles simultaneously?

Why is it a hard problem?

μm

- complex aperiodic pattern
- High material contrast
- Rapidly varying (λ) , big

Very hard forward problem

Even if design is *given*, simulating it requires a super computer for *one* brute-force simulation

Inverse problem intractable?

We are not given the design! Ex: maximize I^2 at focal spot \rightarrow search 10⁶ parameters for best focus



[Arbabi, A. et al. *Nat. Nanotechnol.* **10**, 937–943 (2015)]

Large-scale metasurface optimization by domain decomposition



many possible objective functions, (including broad-band/multi-ω)

focal-spot intensity:

$$f = \left| \int \mathbf{G}_{\mathbf{0}}(\mathbf{r}, \mathbf{r}') \mathbf{J}_{\text{equiv}}(\mathbf{r}') d\mathbf{r}' \right|^2,$$

$$\mathbf{J}_{\text{equiv}} \sim \mathbf{E}_{||}$$

wavefront matching:

$$f = \int |\mathbf{E} - \mathbf{E_0}|^2 \, d\mathbf{r}$$

[Pestourie et al. (2018); Lin et al. (2019)]

- Subdivide surface into small (≤10λ) cells, solve in parallel using either LPA or (better) overlapping non-periodic domains (ODA)
 "Stitch" together using *near-to-farfield transformation* to get fields anywhere.
- Optimize cells (*together*) for any desired objective.

Important Notes:

- You cannot optimize each cell individually. All the DOFs (>10⁶) over the entire surface must be considered and updated together.
- No need to restrict oneself to sub-λ domains; domain >> λ tend to work better.

Few parameters per cell: *library* approach

[R. Pestourie, et al., Optics Express (2018).]



optimized metalens

Monochromatic lens at an angle (focal length = 15000 nm, wavelength = 532 nm, angle = 5 degrees)



A small metalens optimization problem

Monochromatic lens at an angle (focal length = 15000 nm, wavelength = 532 nm, angle = 5 degrees)



Maximize the 0th-order transmitted |E|² at a focal point as a function of pillar widths in every cell (here, 40 pillars).

"Boring" off-the shelf nonlinear optimization algorithm: CCSA algorithm [Svanberg (2001)]

[R. Pestourie, et al., Optics Express (2018).]
Brute-force (FDFD) validation

180,000x faster

(0th-order)



Seconds on a laptop

[R. Pestourie, et al., Optics Express (2018).]

Optimization + experiment: extended depth-of-focus metalens

[collaboration with A. Majumdar, UW (2019)]

e-beam cylindrical lens



(SiN on fused silica)



 44μ m depth of focus, focal length 133μ m

Topology Optimization for Metasurfaces

[Lin et al. (2019)]

Structural evolution of a large-area (100 x 100 λ^2) metalens during topology optimization ~ 10⁶ DOF



Every "pixel" is a degree of freedom ... possibly in multiple layers!

domain decomposition (LPA / ODA) + 1000s of parameters per domain = millions of parameters in total

Difficult metasurface designs



RGB 2-layer lens (NA = $0.71, 200\lambda$ diameter, 50% efficiency)





Concentrator: multiple angles, same focus

now fully achromatic lens...

Achromatic (480–700nm) metalens

[Lin et al (2019)]

Topology-optimization thrives in a large design space ...



15 layers of 140 nm thickTiO₂ NA = 0.71 Lens size: 200 λ s Average focusing efficiency > 50%

Proof-of-concept 2D design: large size + high NA + broadband + by far, the best efficiency

Note: Full FDTD validation of the entire lens. Optimization is not just throwing parameters at a computer.

To get a tractable problem, domain-specific expertise goes into how you formulate the objective & parameters. Many physical similar choices that have very different mathematical properties!

Many design problems remain to be attacked, & several recent bounds far from attained.