## Photonic Inverse Design using the Adjoint Method

Adam Reid - Co-founder and VP Engineering

Lumerical Inc.
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## Photonic Inverse Design Using the Adjoint Method

+ Lumopt ${ }^{1}$ Python module for adjoint sensitivity analysis
+ FDTD Solutions for 2D/3D simulation
+ SciPy gradient based optimization algorithms
= Highly efficient optimization of photonic components

Figure of Merit


Try yourself: Examples and software lumeri.ca/ofc


## Motivation

- Component design challenging, even for basic components
- We would like a lot:
- No reflections
- No loss
- Insensitive to manufacturing imperfections
- Works for range of wavelengths
- Works at different temperatures
- Usually no analytic solution
- Good solutions using PSO
- Zhang, Y., Yang, S., Eu-Jin Lim, A., Lo, G-Q., Galland, C., Baehr-Jones, T., and Hochberg, M., "A compact and low loss Y-junction for submicron silicon waveguide," Optics Express 21, 1310-1316 (2013).
- Can we do better with adjoint methods?



## Lumerical's Suite of Simulation Tools for Photonics

| Optical Simulation | Multiphysics |  |
| :---: | :---: | :---: |
| FDTD Solutions <br> Nanophotonic Design Environment | DEVICE品 <br> Multiphysics Photonics Design Platform |  |
|  | Charge Transport Solver | System El |
|  | Discontinuous Galerkin Time-Domain Solver | Lasere Elen |
| MODE Solutions | Heat Transport Solver | EDA Inter |
| Waveguide Desion Envionment | Finite Element Eigenmode Solver | Foundry |
| Data Exchange: Lumerical Scripting \| MATLAB API |  | hon API |

This demo uses FDTD simulation automated via Python API

## Lumopt: Python Based Inverse Design for Lumerical FDTD

- Lumopt: open source adjoint sensitivity analysis
- Collaboration with Lumerical over past year
- Targets integrated photonics - Now included with FDTD Solutions


## https://github.com/chriskeraly/lumopt

〈>Code (1) Issues 0 \& P Pull requests 2 III Projects 0 国 Wiki Llı Insights
Python based continuous adjoint optimization wrapper for Lumerical

## Adjoint shape optimization applied to electromagnetic design

Christopher M. Lalau-Keraly, ${ }^{1, *}$ Samarth Bhargava, ${ }^{1}$ Owen D. Miller, ${ }^{2}$ and Eli Yablonovitch ${ }^{1}$
${ }^{l}$ Department of Electrical Engineering and Computer Sciences, University of California at Berkeley, Berkeley, California 94720, USA
${ }^{2}$ Department of Mathematics, Massachusetts Institute of Technology, Cambridge, Massachusetts 02139, USA chrisker@eecs.berkeley.edu

Optics Express, Vol 21, Issue 18, 2013
https://www.osapublishing.org/oe/abstract.cfm?uri=oe-21-18-21693

## Inverse Design vs Forward Design

3dB Power Splitter


## Parametric Shape based adjoint optimization

## Parametric shape

- Defines design space
- Optimization parameters

Adjoint sensitivity analysis

- Efficiently compute gradient
- 2 FDTD simulations
- Independent of \# parameters

Gradient based optimization

- Highly efficient optimization
- Uses more physics of device



## Example: Full component design flow for Y-Branch

## Full component design flow for Y-Branch

- Objective: build a splitter like prior art below
- Use inverse design to build splitter section (1)
- Add waveguide offset arms (2) post-optimization

https://github.com/lukasc-ubc/SiEPIC_EBeam_PDK

A compact and low loss Y-junction for submicron silicon waveguide
Yi Zhang, et al, Optics Express Vol. 21, Issue 1, pp. 1310-1316 (2013)
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## An Inverse Design Flow



## Step 1: Define base simulation

2D simulation

- Uses an effective index for waveguide
- Good approximation
- Fast simulation


Base simulation is defined by Lumerical Script (Isf)

## Step 2: Define parametric shape

- Parametric shape defined as Python function
- Function argument is list of parameters
- Function returns list of polygon vertices


```
def taper_splitter(params = np.linspace(0.25e-6, 2e-6, 20))
    Defines a taper where the paramaters are the y coordinates of the nodes of a cubic
    points_x = np.concatenate(([-2.51e-6], np.linspace(-2.5e-6,2.5e-6,20), [2.51e-6]))
    points_y = np.concatenate(([0.25e-6], params, [2e-6]))
    n_interpolation_points = 100
    px = np.linspace(min(points_x), max(points_x), n_interpolation_points)
    interpolator = sp.interpolate.interp1d(points_x, points_y, kind = 'cubic')
    py = interpolator(px)
    py = np.minimum(2.5e-6, py)
    py = np.maximum(np.concatenate((np.ones(50)*0.2e-6, np.ones(50)*0.53e-6)), py)
    px = np.concatenate((px, px[40::][::-1]))
    py = np.concatenate((py, py[40::][::-1]-0.5e-6))
    polygon_points_up = [(x, y) for x, y in zip(px, py)]
    polygon_points_down = [(x, -y) for x, y in zip(px, py)]
    polygon_points = np.array(polygon_points_up[::-1] + polygon_points_down)
    return polygon_points
```


## Step 3: Run fast 2D optimization

This optimization runs in 20-30 minutes






## Step 4: Refine with 3D optimization

- This step is largely the same as 2D simulation
- Takes a bit longer to run
- Should complete with few iterations if seeded with 2D solution



## Step 5: Save design to GDSII

- Optimized shape and output arms saved to GDSII
- Similar to prior art, but has a few ripples!

Example design
Prior art


## Step 5: Y splitter example: Compact model extraction from layout

- Import the final GDSII mask into 3D simulation
- Define ports
- Extract the S-parameters
- Save to data file for INTERCONNECT circuit simulation


Example: Broadband \& Compact Y-Branch

## Broadband \& Compact Splitter

- Can we make a smaller splitter?
- Can we ensure broadband?
https://github.com/lukasc-ubc/SiEPIC_EBeam_PDK
- Parametric shape with output waveguides, 20 parameters
- $5 \times 5$ footprint footprint
- FOM taken over C+L bands



## Broadband Inverse Design

- Optimize FOM over a spectrum
- No additional FDTD simulations required!




## Step 3: Run fast 2D optimization

This example takes $\sim 60$ minutes to run:

$\mathrm{FOM}=0.5$ = ideal

$$
\text { FOM }=0.5 \text { = ideal }
$$

Figure of Merit




Example: Grating coupler

## Available soon!!

Double etch grating coupler
>80 optimization parameters


Example: Robust Y-Branch

## Co-Optimization

Co-optimization:

- Run multiple optimizations concurrently
- Optimizations share same parameters
- Figure of merit or structure can be different


## Example uses:

- Dual polarization devices (different FOM)
- Multiple wavelengths (different FOM)
- Optimize process corners (different geometry)


## Dual polarization co-optimization



## Co-optimization: Robust splitter

- Build a splitter tolerant to manufacturing error
- Co-optimize 2 different shapes (same parameters)
- "Over etch" slightly smaller than nominal
- "Under etch" slightly larger than nominal
- Same FOM function

Over etch


Under etch


## Co-optimization: Robust splitter

- Setup 2 optimizations
- Sum the figures of merit
- 2 FDTD simulations/FOM/iteration


Co-optimization of $+/-20 \mathrm{~nm}$ on edge position


Nice smooth shape!

Upcoming features

## Layout using Cadence's Virtuoso CurvyCore Technology

- Non-Manhattan shapes
- Symbolic equations provide accurate mathematical model
- Generates high-quality polygon representations for fabrication
- Ideal for inverse design shapes

Two Approaches to Inverse Design

Parametric Geometry Optimization

- Finds optimal parameters fo shapes)
- Parametric shape define P lImits design


Topology Optimization

- User provides footprint and 2 materials
- No intuition about shaperrequiryd!
- Solver finds best simon



## Topology Optimization

Supports:

- Broadband
- Quasi-2D
- Constrained feature size
- Co-optimization $\square$


## Topology Optimization: Broadband (1450-1650nm) TE Splitter




## Next Steps

Try running examples

- Stick to 2D, get results in minutes
- Set max_iter=3 to get suboptimal device fast

Try some modifications

- Change device footprint
- Change bandwidth
- Change number of optimization parameters
- Try pCell suggestions in tutorial

More examples available in applications gallery


See more
Exhibit Hall Booth 5438

