

Inverse Design in Photonics

Tutorial I: Introduction



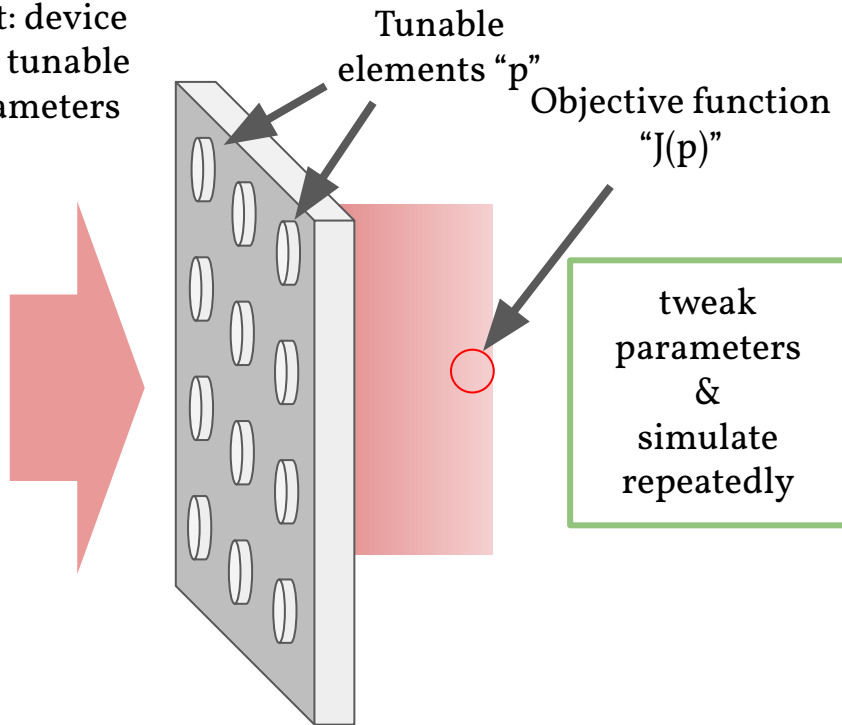
FLEXCOMPUTE



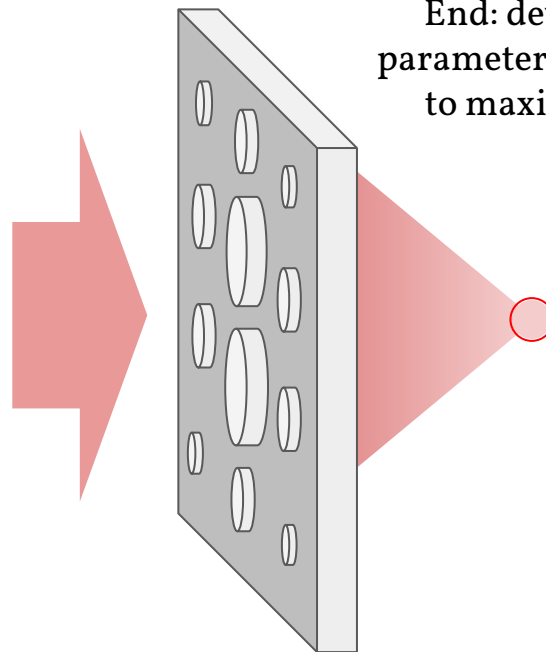


Introduction to Computational Design

Start: device with tunable parameters



End: device with parameters optimized to maximize $J(p)$



[1] Hughes, Tyler W., et al. *Applied Physics Letters* 119.15 (2021): 150502.

[2] Mansouree, Mahdad, et al. *ACS Photonics* 8.2 (2021): 455-463.

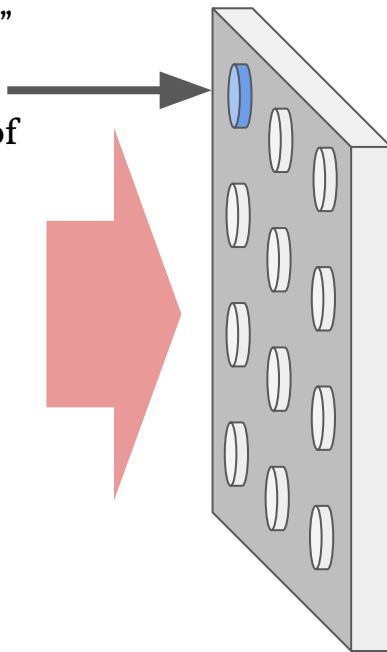


Sensitivity Analysis

Goal: compute
sensitivity " dJ/dp "

Tunable
element " p "

eg. radius of
cylinder

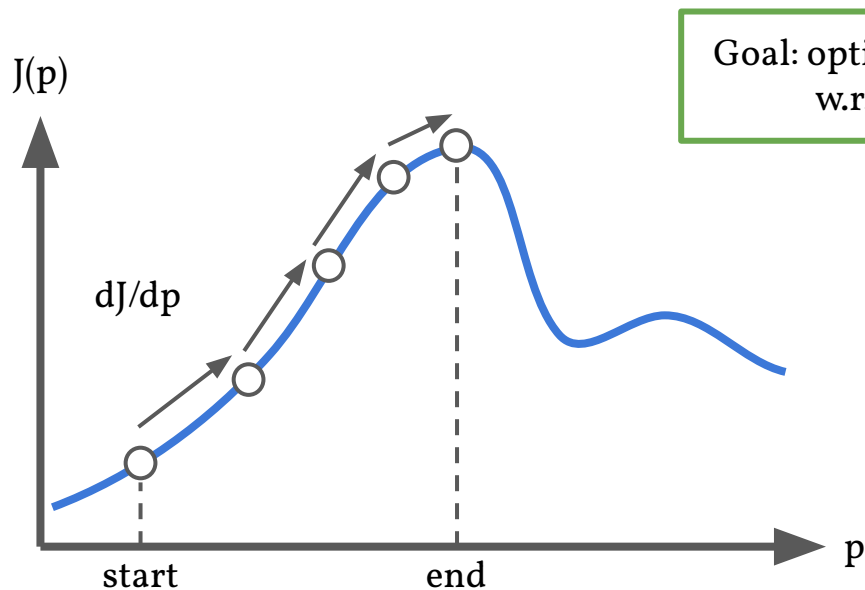


Objective function
" $J(p)$ "

eg. focusing
strength



Optimization (single element)



Goal: optimize $J(p)$
w.r.t p

Approach:

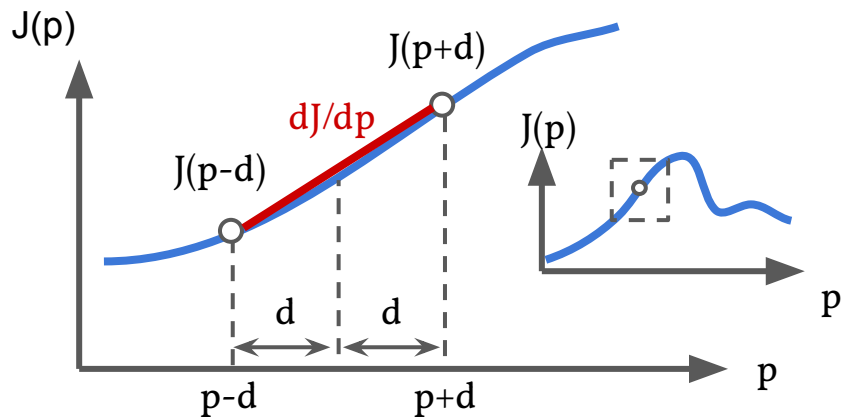
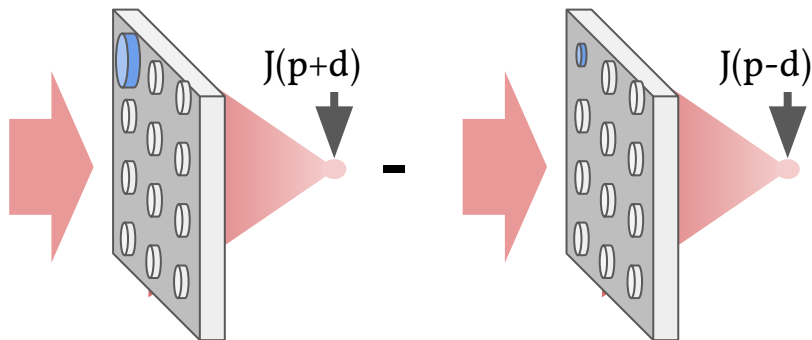
- Start with “ p ”
- Compute dJ/dp
- Adjust p in the direction of dJ/dp
- Repeat until $dJ/dp = 0$



Finite Difference Derivatives

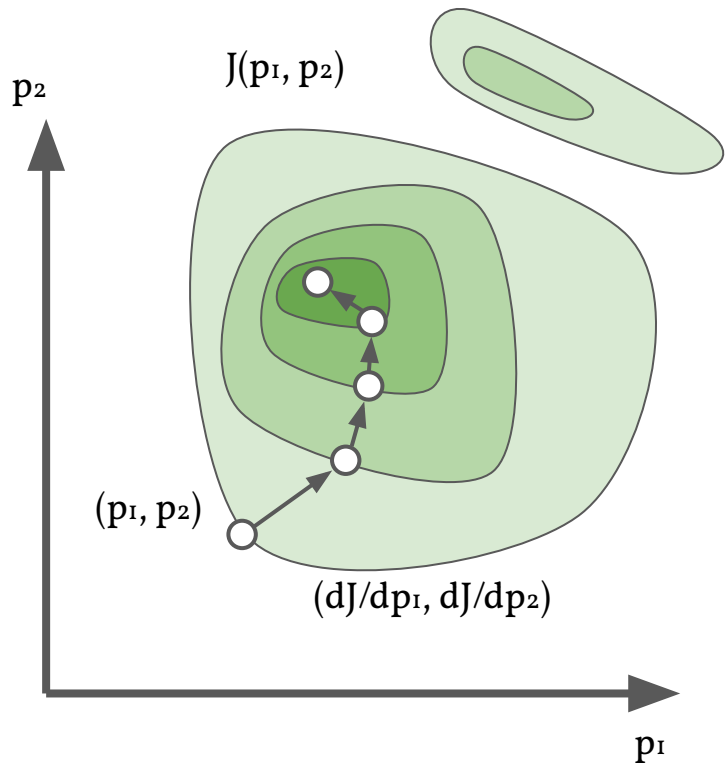
Goal: compute dJ/dp

$$\frac{dJ}{dp} \approx \frac{J(p+d) - J(p-d)}{2d}$$





Multi-Dimensional Sensitivity



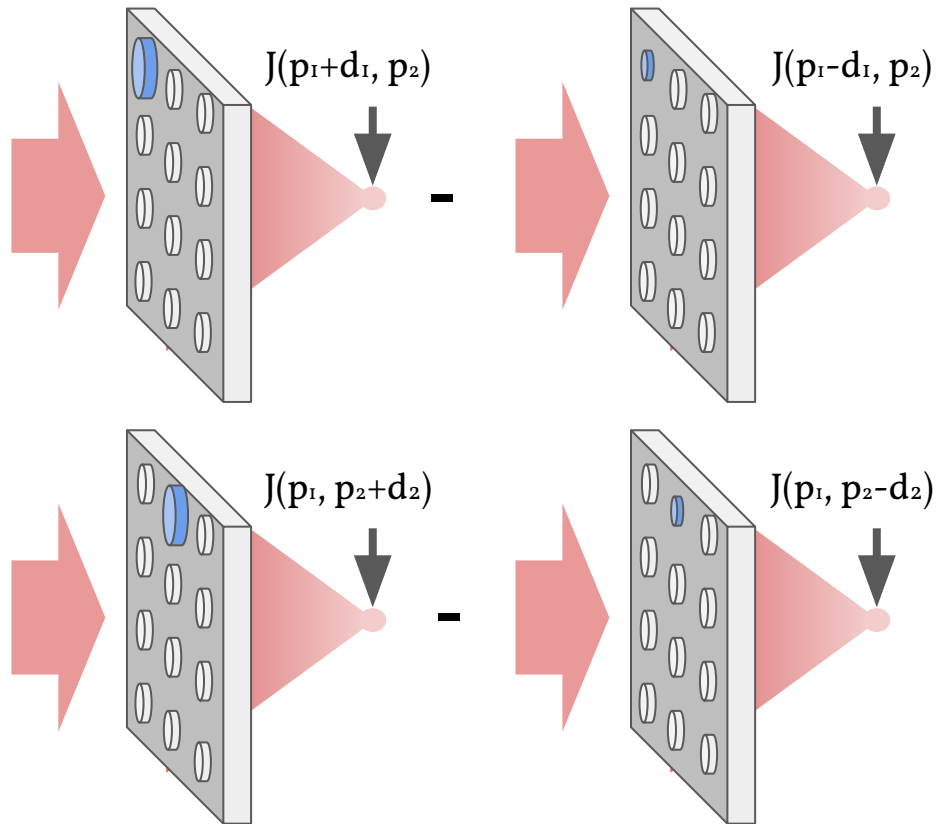
Goal: optimize $J(p_1, p_2)$

Approach (similar to 1D):

- Start with “ (p_1, p_2) ”
- Compute derivatives $(dJ/dp_1, dJ/dp_2)$
- Adjust p_1 in the direction of dJ/dp_1
- Adjust p_2 in the direction of dJ/dp_2
- Repeat until $dJ/dp_1 = 0$ and $dJ/dp_2 = 0$



2D Finite-Difference Derivatives



Goal: compute $(dJ/dp_1, dJ/dp_2)$

$$\frac{dJ}{dp_1} \approx \frac{J(p_1 + d_1, p_2) - J(p_1 - d_1, p_2)}{2d_1}$$

Note: need 4 simulations.

$$\frac{dJ}{dp_2} \approx \frac{J(p_1, p_2 + d_2) - J(p_1, p_2 - d_2)}{2d_2}$$



N-Dimensional Derivative Calculation

Vector Notation:

$$\mathbf{p} = (p_1, p_2, p_3, \dots, p_N)$$

$$J = J(\mathbf{p})$$

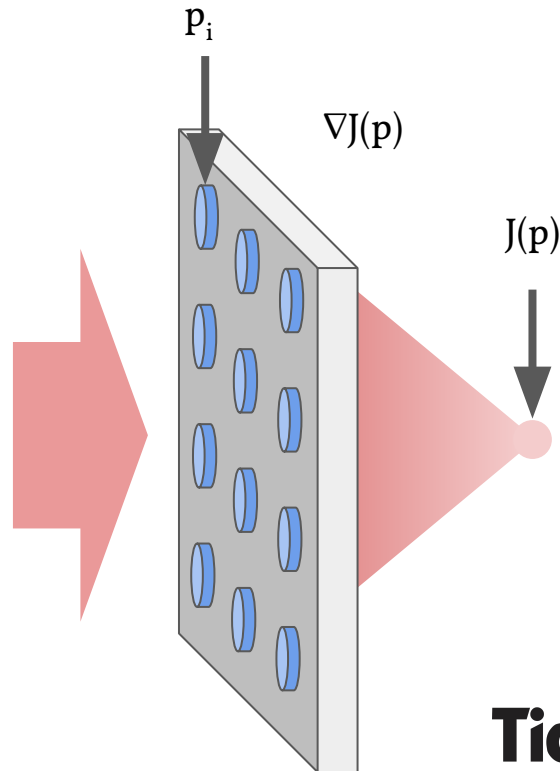
$$\nabla_{\mathbf{p}} J = (dJ/p_1, dJ/p_2, dJ/p_3, \dots, dJ/p_N) = \text{the “gradient”}$$

Computing $\nabla_{\mathbf{p}} J$ through finite difference:

$$\nabla_{\mathbf{p}} J \approx \frac{J(\mathbf{p} + \mathbf{d}_i) - J(\mathbf{p} - \mathbf{d}_i)}{2|\mathbf{d}_i|}$$

Repeat for all N parameters.

- Requires 2N simulations!





Issues with Finite Difference

- In practice, there are significant issues with finite difference:
 - For inverse design, $N \sim$ thousands / millions.
 - Requires choosing step size “ d_i ” for each parameter.
 - Derivatives are approximate.
- Alternative approach: *adjoint method* [3,4]
 - requires only 2 simulations, regardless of N .
 - No need to choose step size.
 - Gives exact gradient, generally without approximation.

[3] <https://math.mit.edu/~stevenj/I8.336/adjoint.pdf>

[4] G. Veronis, R. W. Dutton, and S. Fan, Optics Letters 29, 2288 (2004)