

Fundamental Bounds For Volumetric Structures and Their Feasibility

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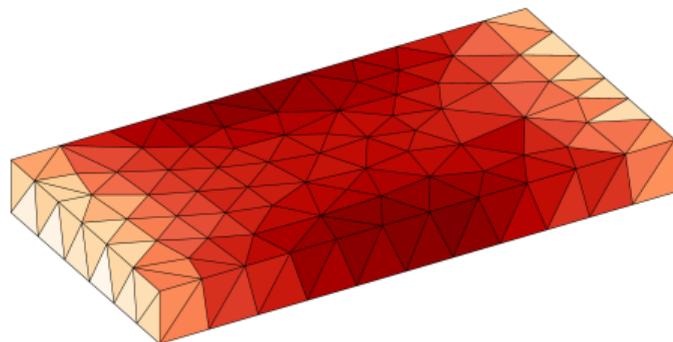
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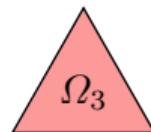
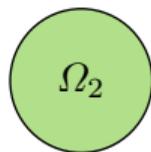
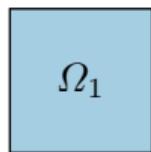
1. Shape Analysis, Synthesis, and Optimal Design
2. Parameterization of a Model
3. Removal and Addition of DOF
4. Topology Sensitivity for VMoM
5. Topology Sensitivity for VMoM: Examples
6. Concluding Remarks



(Sub-)optimal solution of maximum scattering cross section of slab made of gold at 770 MHz.

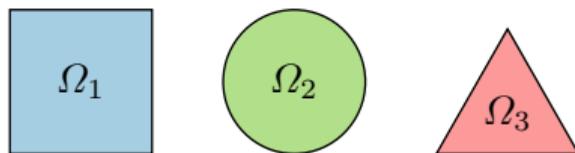


Analysis \times Synthesis





Analysis \times Synthesis



Analysis (\mathcal{A})

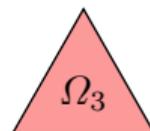
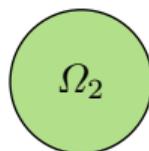
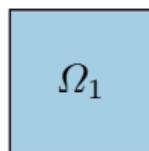
- Shape Ω is given, BCs are known, determine EM quantities.

$$p = \mathcal{L}\mathbf{J}(\mathbf{r}) = \mathcal{A}\{\Omega, \mathbf{E}^i\}$$

- p is an investigated quantity ($Z_{\text{in}}, Q, P_{\text{rad}}, \eta_{\text{rad}}, \dots$) or a composite metric



Analysis \times Synthesis



Analysis (\mathcal{A})

- Shape Ω is given, BCs are known, determine EM quantities.

$$p = \mathcal{L}\mathbf{J}(\mathbf{r}) = \mathcal{A}\{\Omega, \mathbf{E}^i\}$$

Synthesis ($\mathcal{S} \equiv \mathcal{A}^{-1}$)

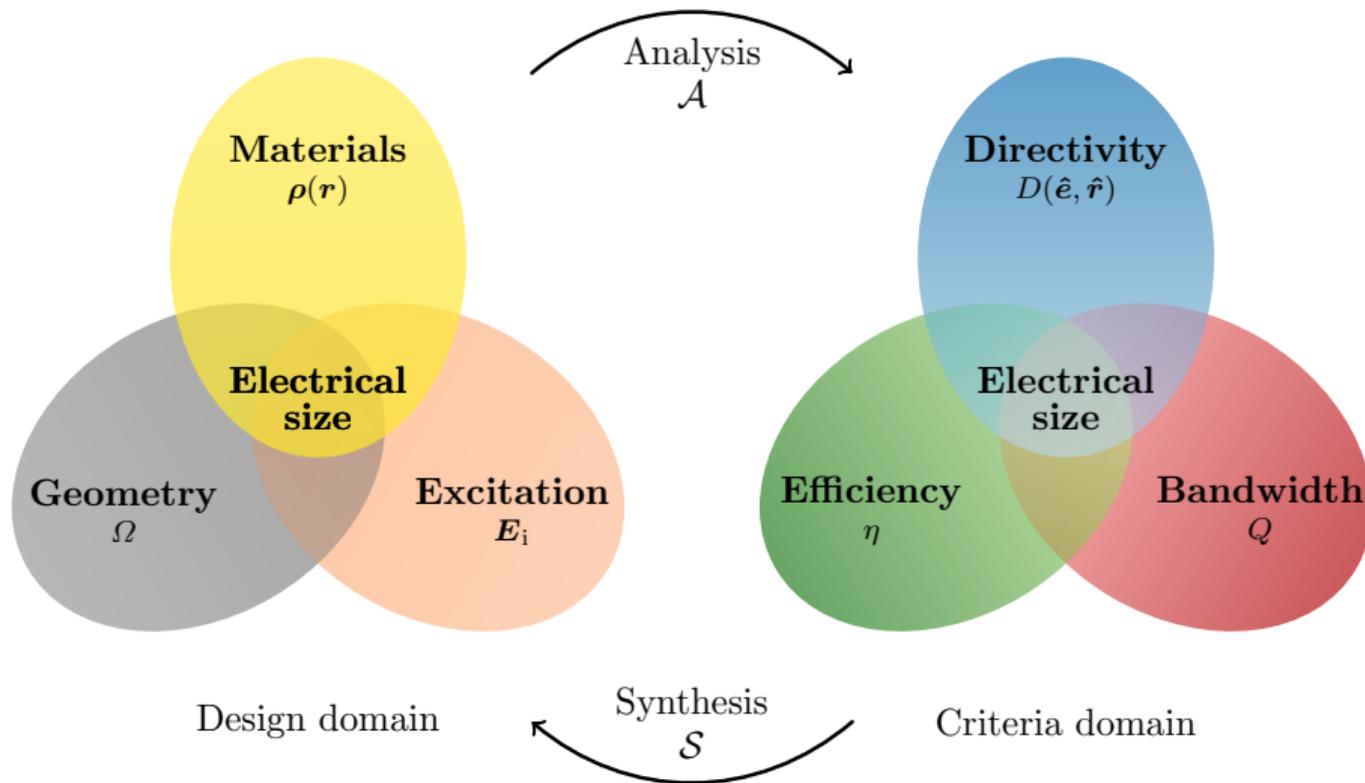
- EM behavior is specified, neither Ω nor BCs are known.

$$\{\Omega, \mathbf{E}^i\} = \mathcal{A}^{-1}p = \mathcal{S}p$$

- p is an investigated quantity ($Z_{\text{in}}, Q, P_{\text{rad}}, \eta_{\text{rad}}, \dots$) or a composite metric



Antenna Design





Shape Synthesis: Rigorous Definition

For a given impedance matrix $\mathbf{Z} \in \mathbb{C}^{N \times N}$, matrices \mathbf{A} , $\{\mathbf{B}_i\}$, $\{\mathbf{B}_j\}$, (a given) excitation vector $\mathbf{V} \in \mathbb{C}^N$, find a vector \mathbf{x} such that

$$\begin{aligned}
 & \text{minimize} && \mathbf{I}^H \mathbf{A}(\mathbf{x}) \mathbf{I} \\
 & \text{subject to} && \mathbf{I}^H \mathbf{B}_i(\mathbf{x}) \mathbf{I} = p_i \\
 & && \mathbf{I}^H \mathbf{B}_j(\mathbf{x}) \mathbf{I} \leq p_j \\
 & && \mathbf{Z}(\mathbf{x}) \mathbf{I} = \mathbf{V} \\
 & && \mathbf{x} \in \{0, 1\}^N
 \end{aligned}$$

$$\mathbf{x} = \begin{bmatrix} 1 & 1 & 0 & \cdots & 1 \\ Z_{11} & Z_{12} & Z_{13} & \cdots & Z_{1N} \\ Z_{21} & Z_{22} & Z_{23} & \cdots & Z_{2N} \\ Z_{31} & Z_{32} & Z_{33} & \cdots & Z_{3N} \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ Z_{N1} & Z_{N2} & Z_{N3} & \cdots & Z_{NN} \end{bmatrix} \begin{bmatrix} I_1 \\ I_2 \\ I_3 \\ \vdots \\ I_N \end{bmatrix} = \begin{bmatrix} V_1 \\ V_2 \\ V_3 \\ \vdots \\ V_N \end{bmatrix}$$

¹G. L. Nemhauser and L. A. Wolsey, *Integer and Combinatorial Optimization*. John Wiley & Sons, 1999, ISBN: 0-471-35943-2



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 \begin{bmatrix} I_1 \\ I_2 \\ I_3 \\ \vdots \\ I_N \end{bmatrix}
 =
 \begin{bmatrix} V_1 \\ V_2 \\ V_3 \\ \vdots \\ V_N \end{bmatrix}$$

- ▶ Combinatorial optimization¹ (suffers from **curse of dimensionality**, 2^N possible solutions),
- ▶ vector \mathbf{x} serves as a characteristic function (structure perturbation).

¹G. L. Nemhauser and L. A. Wolsey, *Integer and Combinatorial Optimization*. John Wiley & Sons, 1999, ISBN: 0-471-35943-2



Degrees of Freedom

Bounding box

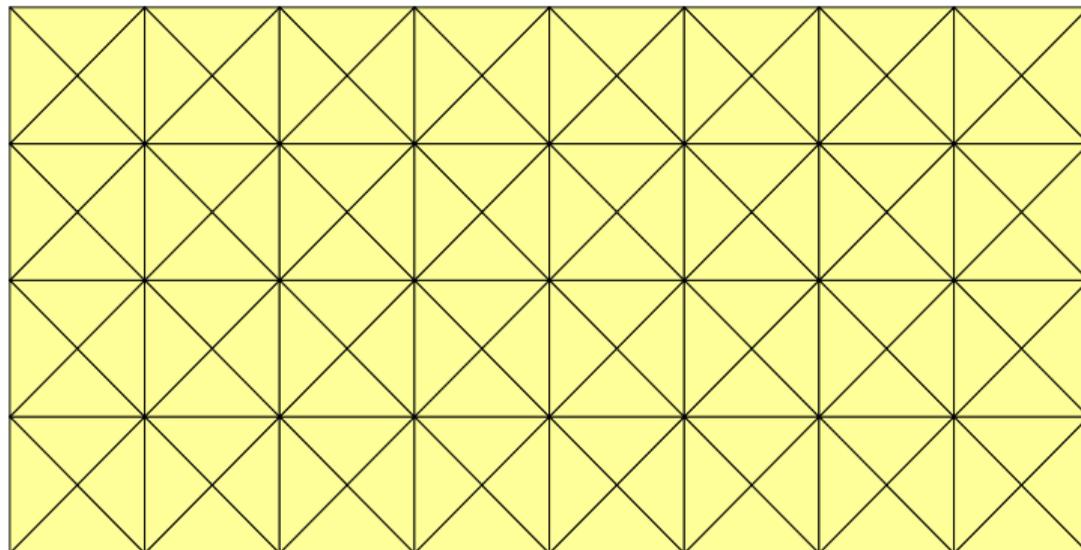
 Ω 



Degrees of Freedom

Bounding box \rightarrow discretization

$$\Omega \rightarrow \{T_t\}$$

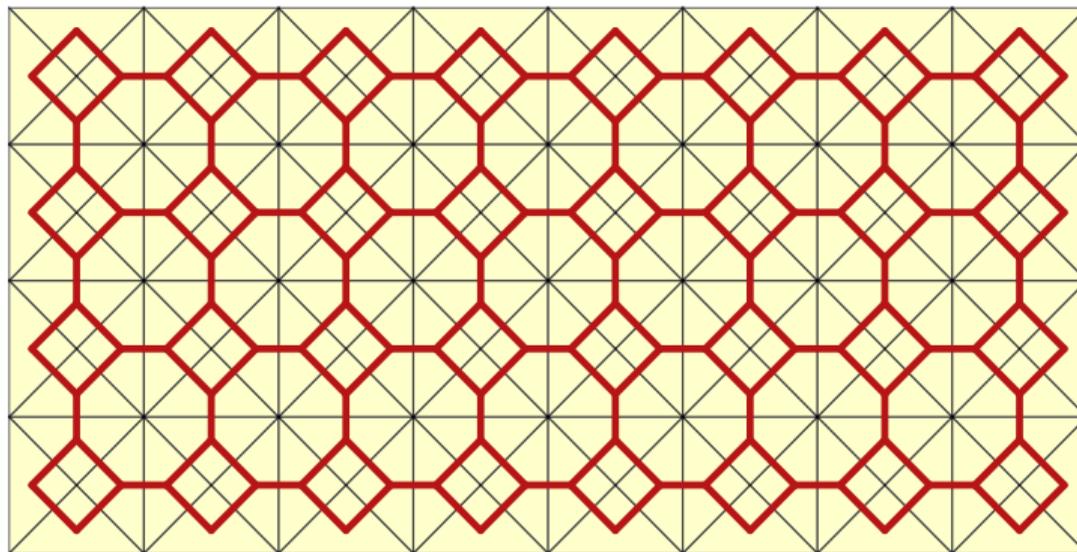




Degrees of Freedom

Bounding box \rightarrow discretization \rightarrow basis functions

$$\Omega \rightarrow \{T_t\} \rightarrow \{\psi_n(\mathbf{r})\}$$

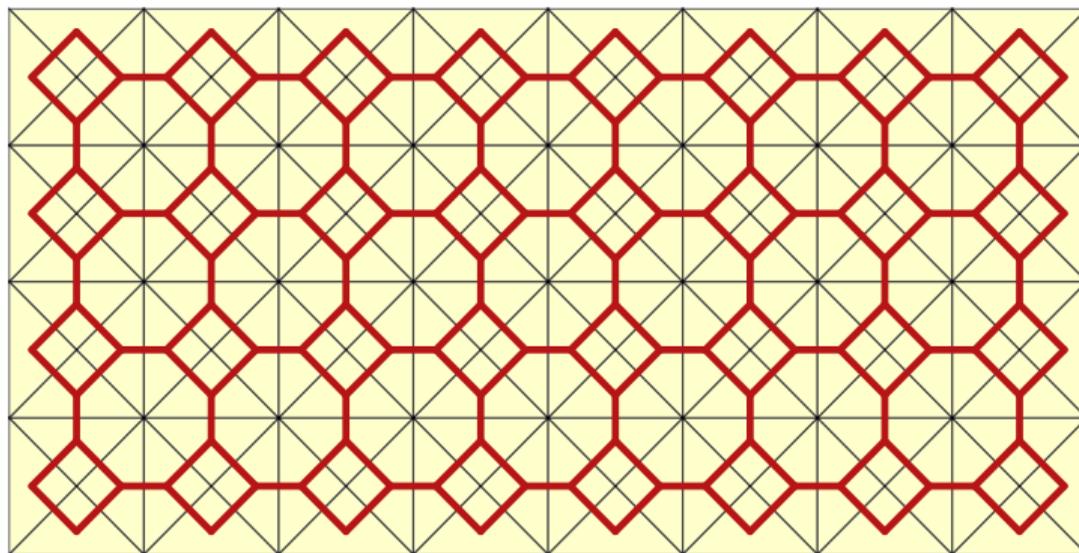




Degrees of Freedom

Bounding box \rightarrow discretization \rightarrow basis functions \rightarrow degrees of freedom to be optimized

$$\Omega \rightarrow \{T_t\} \rightarrow \{\psi_n(\mathbf{r})\} \rightarrow \mathbf{g}$$



- ▶ $\mathbf{g} \in \{0, 1\}^{N \times 1}$ is characteristic vector (discretized characteristic function)

Shape Optimization With Exact Reanalysis



Capability to effectively remove/add a degree of freedom.²

²M. Capek, L. Jelinek, and M. Gustafsson, “Shape synthesis based on topology sensitivity,” *IEEE Trans. Antennas Propag.*, vol. 67, no. 6, pp. 3889–3901, 2019. DOI: [10.1109/TAP.2019.2902749](https://doi.org/10.1109/TAP.2019.2902749)



Shape Optimization With Exact Reanalysis

Capability to effectively remove/add a degree of freedom.²

- ▶ Perfectly compatible with method of moments;
 - ▶ basis functions used as DOF.

Example of topology sensitivity, $ka = 1/2$, plate fed in the middle.

²M. Capek, L. Jelinek, and M. Gustafsson, “Shape synthesis based on topology sensitivity,” *IEEE Trans. Antennas Propag.*, vol. 67, no. 6, pp. 3889–3901, 2019. DOI: [10.1109/TAP.2019.2902749](https://doi.org/10.1109/TAP.2019.2902749)



Shape Optimization With Exact Reanalysis

Capability to effectively remove/add a degree of freedom.²

- ▶ Perfectly compatible with method of moments;
 - ▶ basis functions used as DOF.
- ▶ Inversion-free for the smallest perturbations;
 - ▶ gradient-based shape optimization possible deterministically.

Example of topology sensitivity, $ka = 1/2$, plate fed in the middle.

²M. Capek, L. Jelinek, and M. Gustafsson, “Shape synthesis based on topology sensitivity,” *IEEE Trans. Antennas Propag.*, vol. 67, no. 6, pp. 3889–3901, 2019. DOI: [10.1109/TAP.2019.2902749](https://doi.org/10.1109/TAP.2019.2902749)



Update of an EM System With Exact Reanalysis

Modification of the shape described by impedance matrix \mathbf{Z}

$$\mathbf{Y} = \mathbf{Z}^{-1} = \begin{bmatrix} \mathbf{A} & \mathbf{B} \\ \mathbf{C} & \mathbf{D} \end{bmatrix} = \begin{bmatrix} \mathbf{A}^{-1} + \mathbf{A}^{-1}\mathbf{B}\mathbf{E}^{-1}\mathbf{C}\mathbf{A}^{-1} & -\mathbf{A}^{-1}\mathbf{B}\mathbf{E}^{-1} \\ -\mathbf{E}^{-1}\mathbf{C}\mathbf{A}^{-1} & \mathbf{E}^{-1} \end{bmatrix}$$

and consequently perturbation of the current density

$$\mathbf{I} = \mathbf{Y}\mathbf{V}. \quad (1)$$



Removing and Adding DOF (Delta Gap Feeder)

DOF removal:

$$\hat{\mathbf{I}} = \left(\mathbf{y}_f - \frac{Y_{fb}}{Y_{bb}} \mathbf{y}_b \right) l_f V_0,$$

Admittance matrix update:

$$\hat{\mathbf{Y}} = \mathbf{C}^T \left(\mathbf{Y} - \frac{1}{Y_{bb}} \mathbf{y}_b \mathbf{y}_b^T \right) \mathbf{C},$$

DOF addition:

$$\hat{\mathbf{I}} = \mathbf{C}^T \left(\begin{bmatrix} \mathbf{y}_f \\ 0 \end{bmatrix} + \frac{x_{fb}}{z_b} \begin{bmatrix} \mathbf{x}_b \\ -1 \end{bmatrix} \right) l_f V_0,$$

Admittance matrix update:

$$\hat{\mathbf{Y}} = \frac{1}{z_b} \mathbf{C}^T \begin{bmatrix} z_b \mathbf{Y} + \mathbf{x}_b \mathbf{x}_b^T & -\mathbf{x}_b \\ -\mathbf{x}_b^T & 1 \end{bmatrix} \mathbf{C},$$

$$C_{nn} = \begin{cases} 0 & \Leftrightarrow g_n = b \\ 1 & \Leftrightarrow \text{otherwise} \end{cases}$$

$$\mathbf{x}_b = \mathbf{Y} \tilde{\mathbf{z}}_b, \quad z_b = \tilde{Z}_{bb} - \tilde{\mathbf{z}}_b^T \mathbf{x}_b$$

$$C_{mn} = \begin{cases} 1 & \Leftrightarrow g_n = S(m) \\ 0 & \Leftrightarrow \text{otherwise} \end{cases}$$

- All columns of \mathbf{C} matrix containing solely zeros are eliminated before use.



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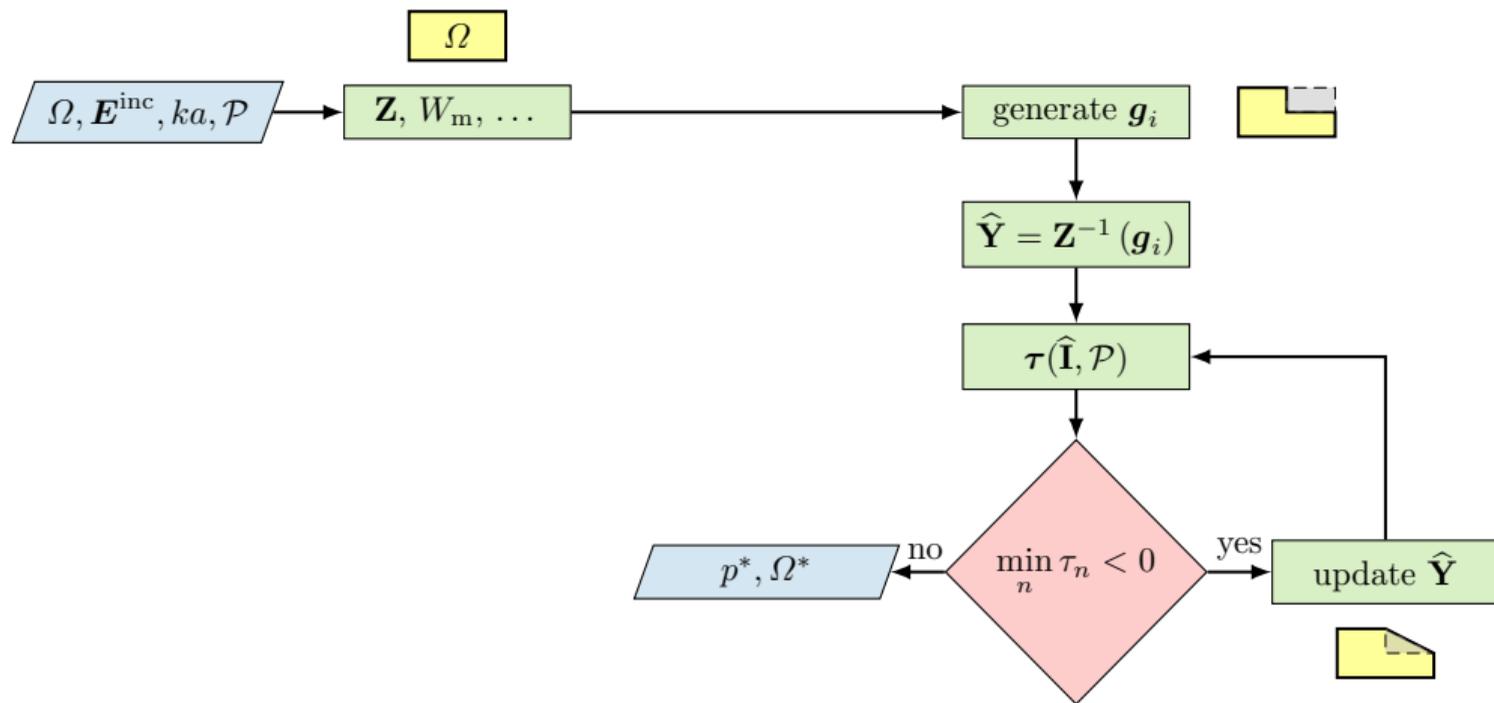
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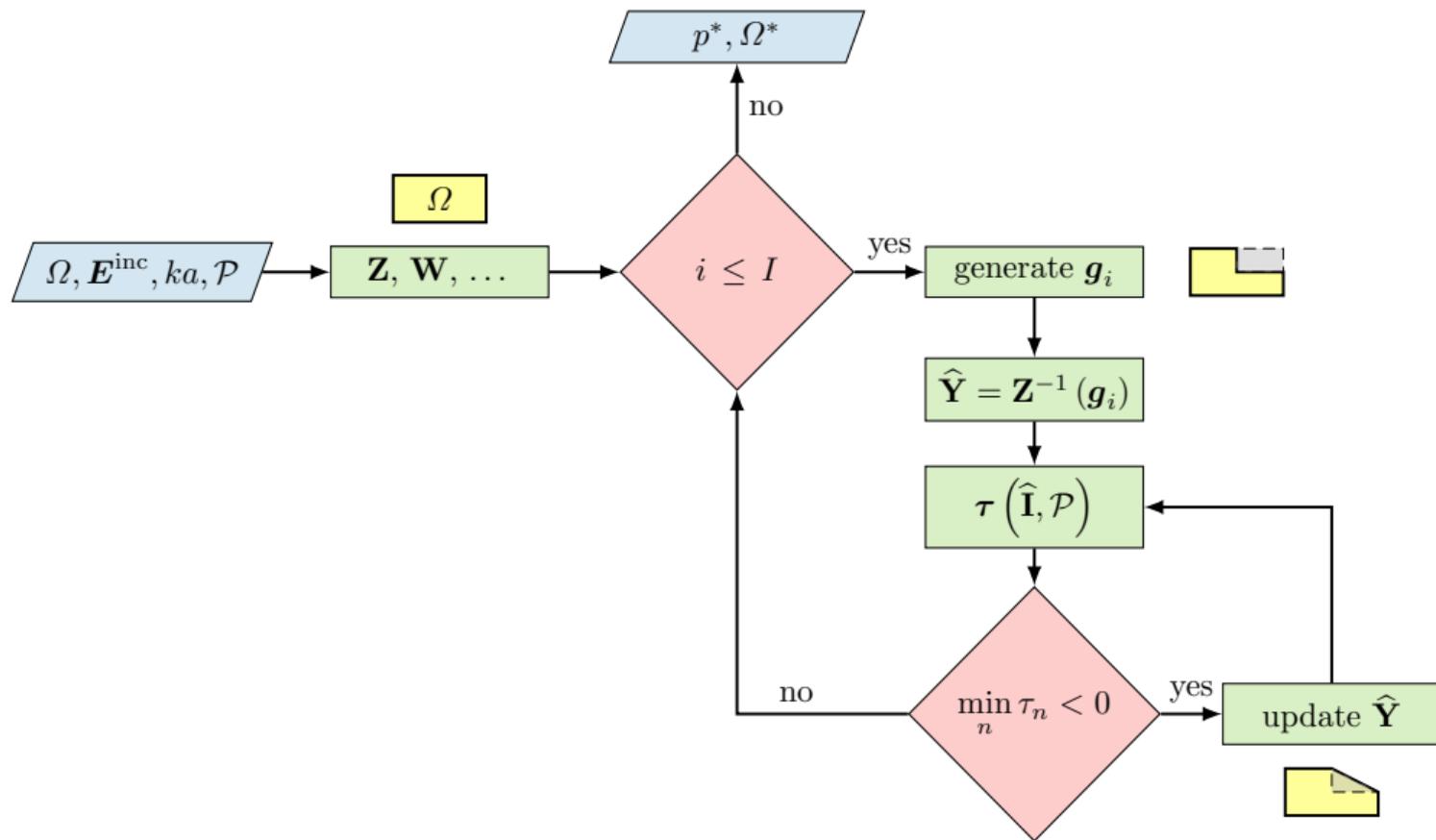
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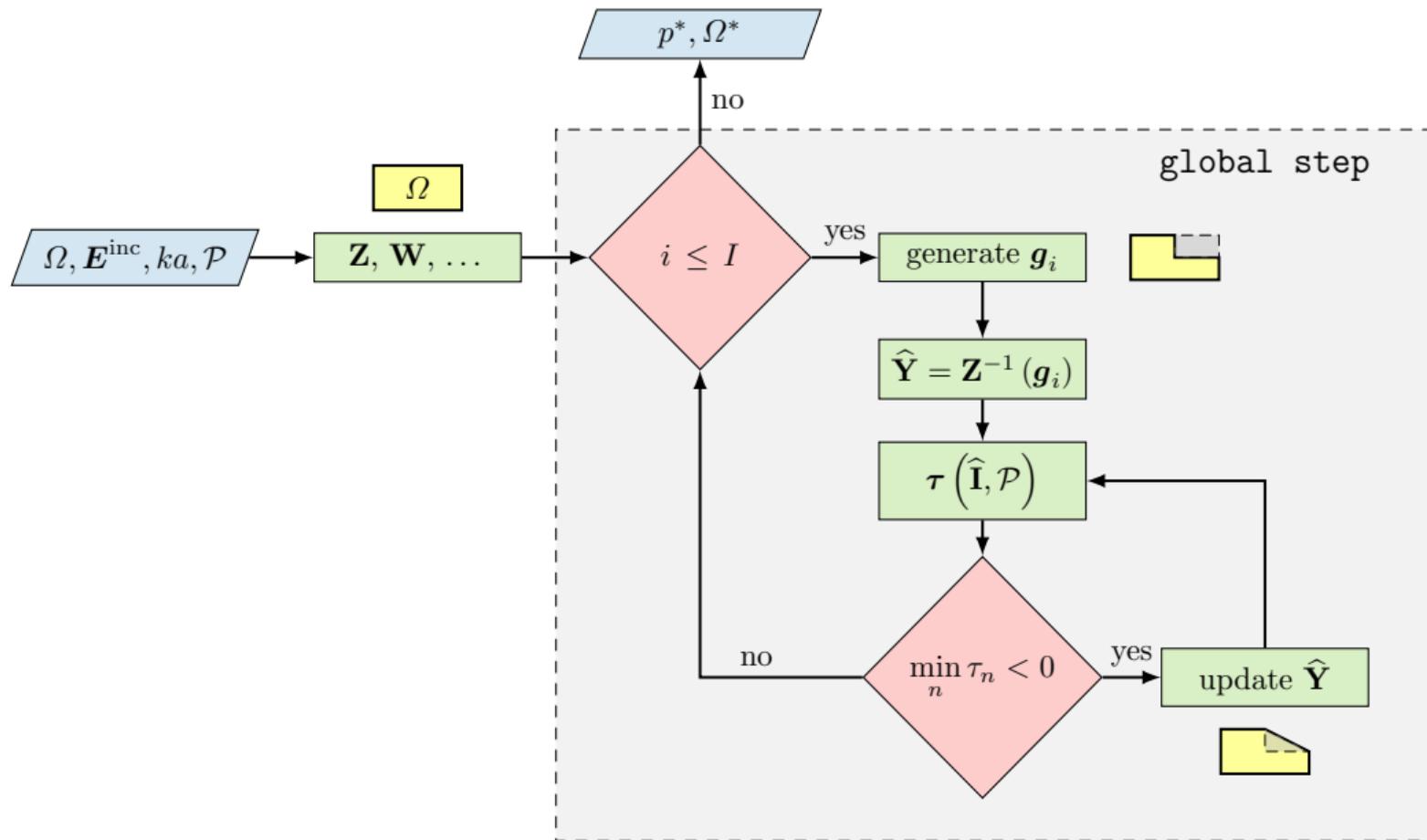
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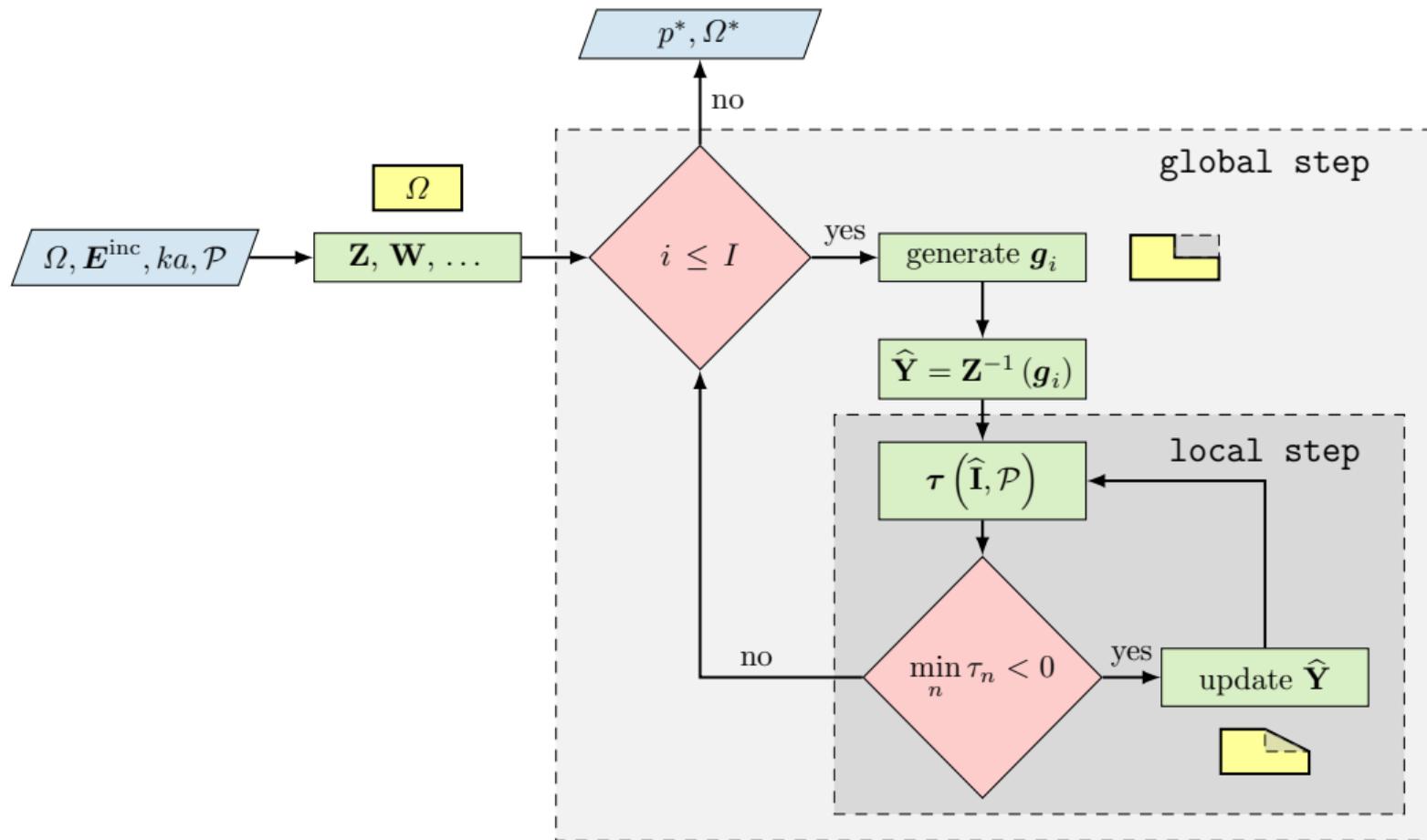
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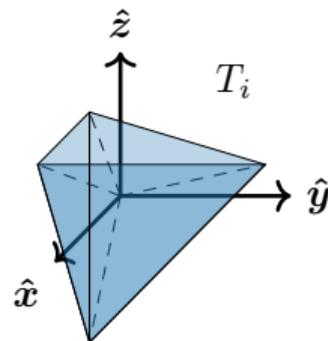






Volumetric Method of Moments

- ▶ Tetrahedral discretization.
- ▶ Piece-wise constant basis functions.
- ▶ (Expensive) volumetric quadrature converted to surface integrals³.



$$Z_{mn} = -j \frac{Z_0}{k} \int_{V_m} \boldsymbol{\psi}_m(\mathbf{r}) \cdot (\mathbf{1} + \boldsymbol{\chi}^{-1}(\mathbf{r})) \cdot \boldsymbol{\psi}_n(\mathbf{r}) dV - j \frac{Z_0}{k} \oint_{S_m} \oint_{S_n} \Psi_{mn}(\mathbf{r}, \mathbf{r}') G(\mathbf{r}, \mathbf{r}') dS' dS,$$

$$\Psi_{mn}(\mathbf{r}, \mathbf{r}') = \boldsymbol{\psi}_m(\mathbf{r}) \cdot \mathbf{n}_n(\mathbf{r}') \boldsymbol{\psi}_n(\mathbf{r}') \cdot \mathbf{n}_m(\mathbf{r}) - \boldsymbol{\psi}_m(\mathbf{r}) \cdot \boldsymbol{\psi}_n(\mathbf{r}') \mathbf{n}_n(\mathbf{r}') \cdot \mathbf{n}_m(\mathbf{r})$$

³A. Polimeridis, J. Villena, L. Daniel, *et al.*, “Stable FFT-JVIE solvers for fast analysis of highly inhomogeneous dielectric objects,” *Journal of Computational Physics*, vol. 269, pp. 280–296, 2014. DOI: [10.1016/j.jcp.2014.03.026](https://doi.org/10.1016/j.jcp.2014.03.026). [Online]. Available: <https://doi.org/10.1016/j.jcp.2014.03.026>



Radiation Efficiency Bounds

- The optimization problems \mathcal{P}_1 and \mathcal{P}_2 can rigorously be formulated.

Maximum radiation efficiency

Problem \mathcal{P}_1 :

$$\begin{aligned} & \text{minimize} && P_{\text{loss}} \\ & \text{subject to} && P_{\text{rad}} = 1 \end{aligned}$$

Maximum self-resonant radiation efficiency

Problem \mathcal{P}_2 :

$$\begin{aligned} & \text{minimize} && P_{\text{loss}} \\ & \text{subject to} && P_{\text{rad}} = 1 \\ & && \omega (W_{\text{m}} - W_{\text{e}}) = 0 \end{aligned}$$



Radiation Efficiency Bounds

- ▶ The optimization problems \mathcal{P}_1 and \mathcal{P}_2 can rigorously be formulated.
- ▶ Having quadratic forms for the physical quantities, the antenna metrics may be optimized.

Maximum radiation efficiency

Problem \mathcal{P}_1 :

$$\begin{aligned} & \text{minimize} && \mathbf{I}^H \mathbf{R}_\rho \mathbf{I} \\ & \text{subject to} && \mathbf{I}^H \mathbf{R}_0 \mathbf{I} = 1 \end{aligned}$$

Maximum self-resonant radiation efficiency

Problem \mathcal{P}_2 :

$$\begin{aligned} & \text{minimize} && \mathbf{I}^H \mathbf{R}_\rho \mathbf{I} \\ & \text{subject to} && \mathbf{I}^H \mathbf{R}_0 \mathbf{I} = 1 \\ & && \mathbf{I}^H \mathbf{X} \mathbf{I} = 0 \end{aligned}$$



Radiation Efficiency Bounds

- ▶ The optimization problems \mathcal{P}_1 and \mathcal{P}_2 can rigorously be formulated.
- ▶ Having quadratic forms for the physical quantities, the antenna metrics may be optimized.
- ▶ The problems \mathcal{P}_1 and \mathcal{P}_2 are quadratically constrained quadratic programs⁴ (QCQP).

Maximum radiation efficiency

Problem \mathcal{P}_1 :

$$\begin{aligned} & \text{minimize} && \mathbf{I}^H \mathbf{R}_\rho \mathbf{I} \\ & \text{subject to} && \mathbf{I}^H \mathbf{R}_0 \mathbf{I} = 1 \end{aligned}$$

Maximum self-resonant radiation efficiency

Problem \mathcal{P}_2 :

$$\begin{aligned} & \text{minimize} && \mathbf{I}^H \mathbf{R}_\rho \mathbf{I} \\ & \text{subject to} && \mathbf{I}^H \mathbf{R}_0 \mathbf{I} = 1 \\ & && \mathbf{I}^H \mathbf{X} \mathbf{I} = 0 \end{aligned}$$



Algebraic Representation of Integral Operators

Radiated and reactive power

Complex power balance:

$$P_{\text{rad}} + P_{\text{lost}} + 2j\omega (W_m - W_e) = \frac{1}{2} \langle \mathbf{J}(\mathbf{r}), \mathcal{Z}[\mathbf{J}(\mathbf{r})] \rangle \approx \frac{1}{2} \mathbf{I}^H \mathbf{Z} \mathbf{I} \quad (2)$$

Radiated power:

$$P_{\text{rad}} \approx \frac{1}{2} \mathbf{I}^H \mathbf{R}_0 \mathbf{I} \quad (3)$$

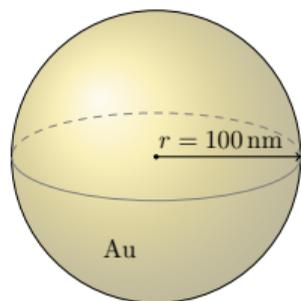
Lost power:

$$P_{\text{lost}} \approx \frac{1}{2} \mathbf{I}^H \mathbf{R}_\rho \mathbf{I} \quad (4)$$

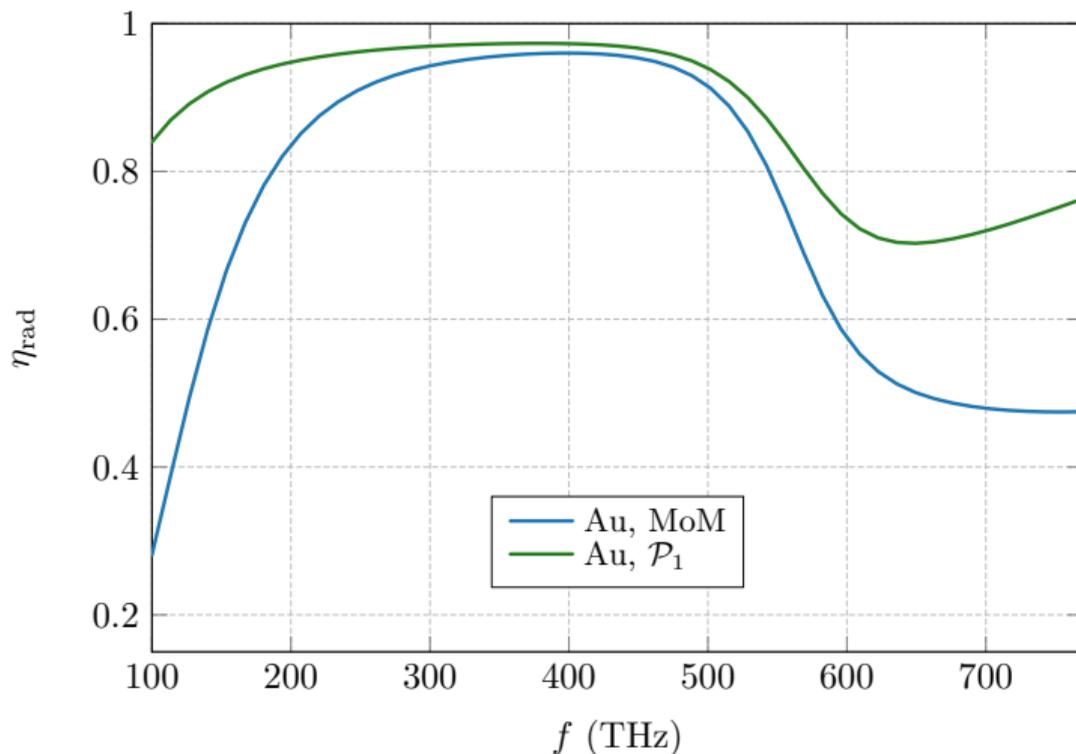
$$Z_{mn} = -j \frac{Z_0}{k} \int_{V_m} \boldsymbol{\psi}_m(\mathbf{r}) \cdot (\mathbf{1} + \boldsymbol{\chi}^{-1}(\mathbf{r})) \cdot \boldsymbol{\psi}_n(\mathbf{r}) dV - j \frac{Z_0}{k} \oint_{S_m} \oint_{S_n} \Psi_{mn}(\mathbf{r}, \mathbf{r}') G(\mathbf{r}, \mathbf{r}') dS' dS,$$



Example: A Gold Nanoparticle – Radiation Efficiency



A particle (Au).



Radiation efficiency: MoM solution compared with optimal performance.



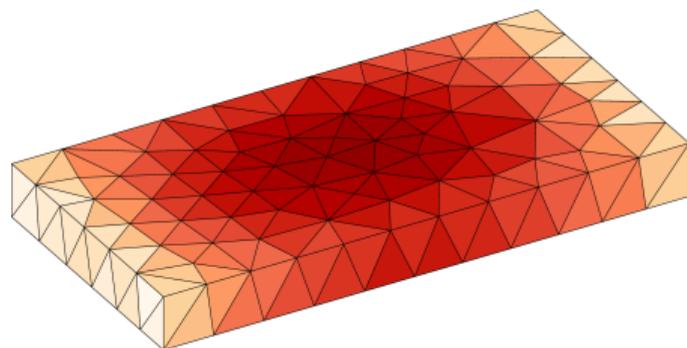
Scattering Cross Section in Dielectric Slab

Fundamental bound⁵:

$$\begin{aligned} &\text{minimize} && \frac{1}{2} \mathbf{I}^H \mathbf{R}_0 \mathbf{I} \\ &\text{subject to} && \mathbf{I}^H \mathbf{R} \mathbf{I} - \text{Re}\{\mathbf{I}^H \mathbf{V}\} = 0 \\ &&& \mathbf{I}^H \mathbf{X} \mathbf{I} - \text{Im}\{\mathbf{I}^H \mathbf{V}\} = 0 \end{aligned}$$

MoM evaluation (for topology optimization):

$$\begin{aligned} \mathbf{I} &= \mathbf{Z}^{-1} \mathbf{V} \\ \sigma_{\text{scat}} &= \frac{1}{2} \frac{\mathbf{I}^H \mathbf{R}_0 \mathbf{I}}{S_0} \end{aligned}$$



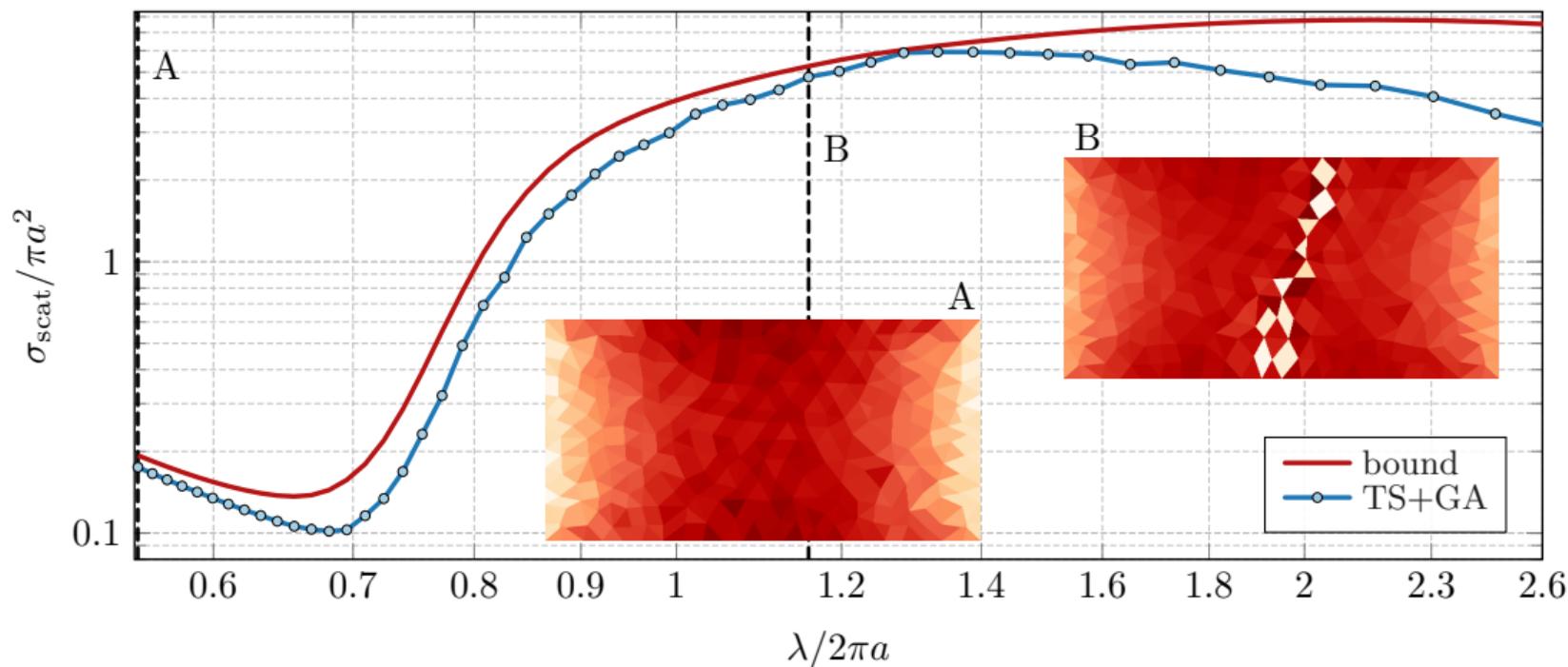
Optimal current maximizing scattering cross section at 770 MHz, VMoM, golden slab.

⁵M. Gustafsson, K. Schab, L. Jelinek, *et al.*, “Upper bounds on absorption and scattering,” 2019, eprint arXiv: 1912.06699. [Online]. Available: <https://arxiv.org/abs/1912.06699>

⁶A. Derkachova, K. Kolwas, and I. Demchenko, “Dielectric function for gold in plasmonics applications: Size dependence of plasmon resonance frequencies and damping rates for nanospheres,” *Plasmonics*, vol. 11, pp. 941–951, 3 Jun. 2016. DOI: [10.1007/s11468-015-0128-7](https://doi.org/10.1007/s11468-015-0128-7)



Comparison of Fundamental Bounds and Optimal Designs



Optimization setting: slab $\ell \times \ell/2 \times \ell/10$, $\ell = 200$ nm, $f \in [160, 770]$ THz, gold⁶, plane wave (\mathbf{V}) polarized along x axis, perpendicular angle of incidence, 1380 basis functions.



Concluding Remarks

What has been done

- ▶ Deterministic inversion-free structure perturbation (removal/addition).
- ▶ A novel memetic algorithm.
- ▶ Knowledge of gradients for a given (fixed) EM model.
- ▶ Robustness and immunity against local minima.



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Topics of ongoing research

- ▶ Acceleration on GPUs.
- ▶ Utilization of big data gathered during the optimization.
- ▶ Regularization to remove irregularities.
- ▶ Adaptive Greedy strategies to overcome slow convergence.
- ▶ How to interpret enabled/disabled $\{\hat{\mathbf{x}}, \hat{\mathbf{y}}, \hat{\mathbf{z}}\}$ basis functions? Anisotropic material?

Questions?

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The presentation is available at

▶ capek.elmag.org

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