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# Successive Iteration of Analysis and Design (SIAD) for large-scale eigenvalue topology optimization

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# Background: Large scale eigenvalue-related TO

- Natural frequency/ forbidden interval/ frequency gap
- PnC and PtC metamaterials
- Linear buckling

**Accurate analysis requires fine FE discretization.**

**Example:** frequency maximization

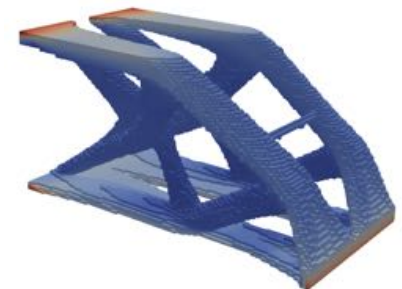
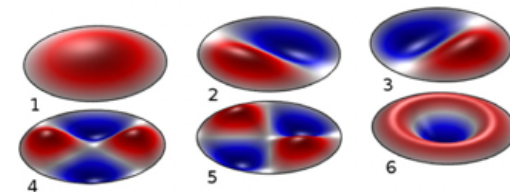
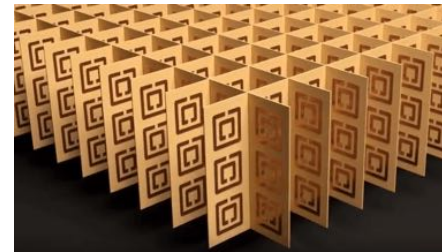
$$\max : \omega_j$$

$$\text{s.t. } \mathbf{K}\boldsymbol{\varphi}_j = \omega_j^2 \mathbf{M}\boldsymbol{\varphi}_j$$

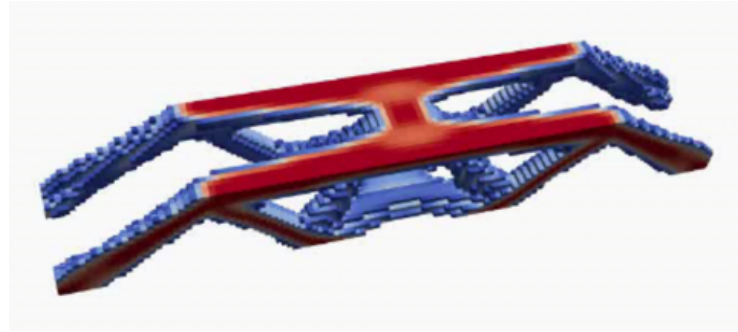
$$\sum_{e=1}^{N_e} x_e V_e - f_v V_0 \leq 0$$

$$0 < \underline{x} \leq x_e \leq 1, \quad e = 1, \dots, N_e$$

$$\text{RAMP model } E_e = \frac{x_e}{1 + p(1 - x_e)} E_0, \quad \rho_e = x_e \rho_0$$



# Major issue: computational cost for analysis



Cost for static compliance minimization of a 12.83 Million DOF model

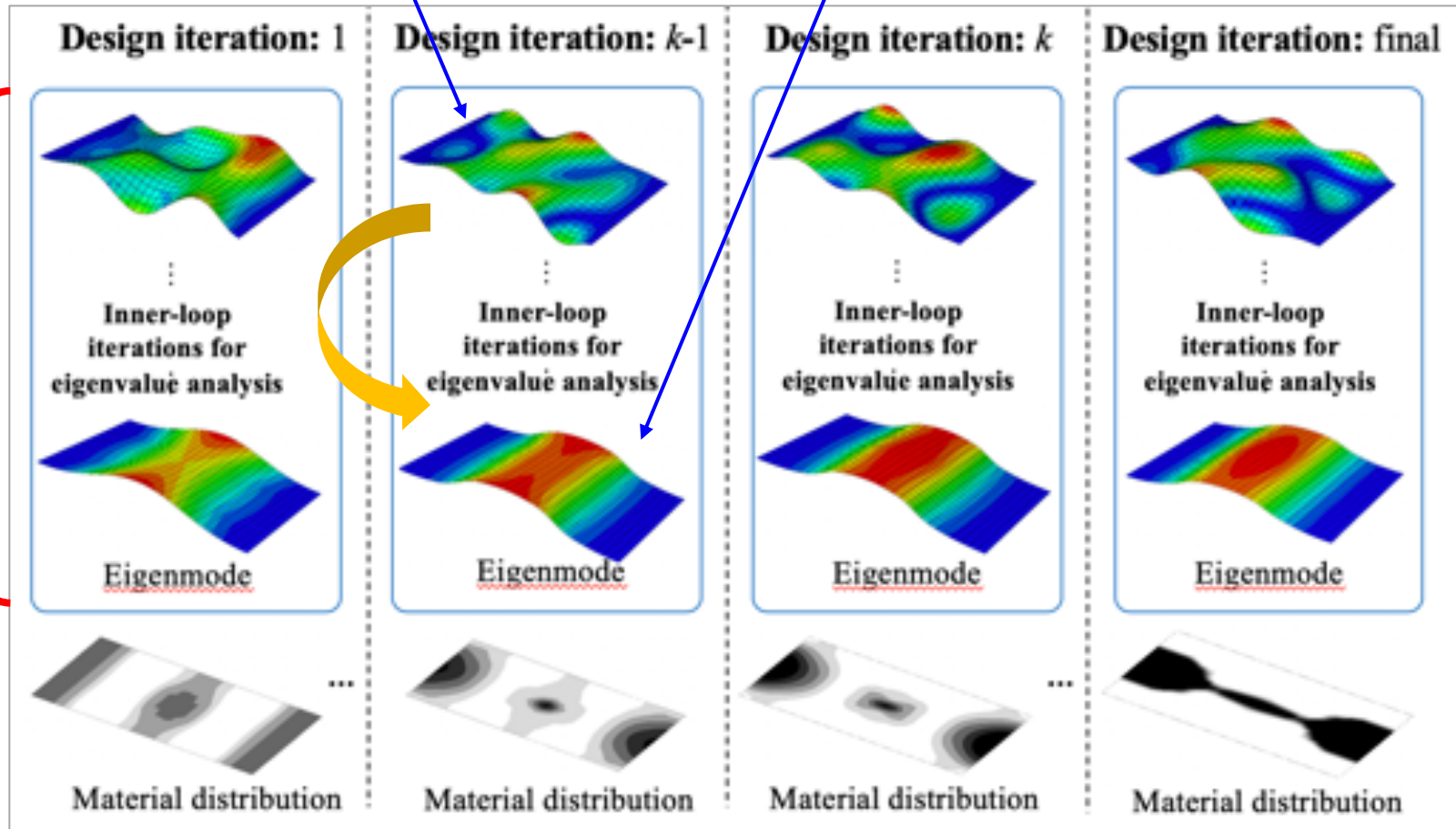
Procedures	Computing time per iteration
Assembling global stiffness matrices	33 sec. (3.1%)
<b>Finite element solution <math>Ku = f</math></b>	<b>1005 sec. (95.5%)</b>
Sensitivity analysis & optimization	14 sec. (1.3%)

For eigenvalue-related problems, the difficulty is even worsened by higher-order eigenvalues, and repeated eigenvalues.

# Conventional (nested double-loop) approach

Initial trial eigenmode

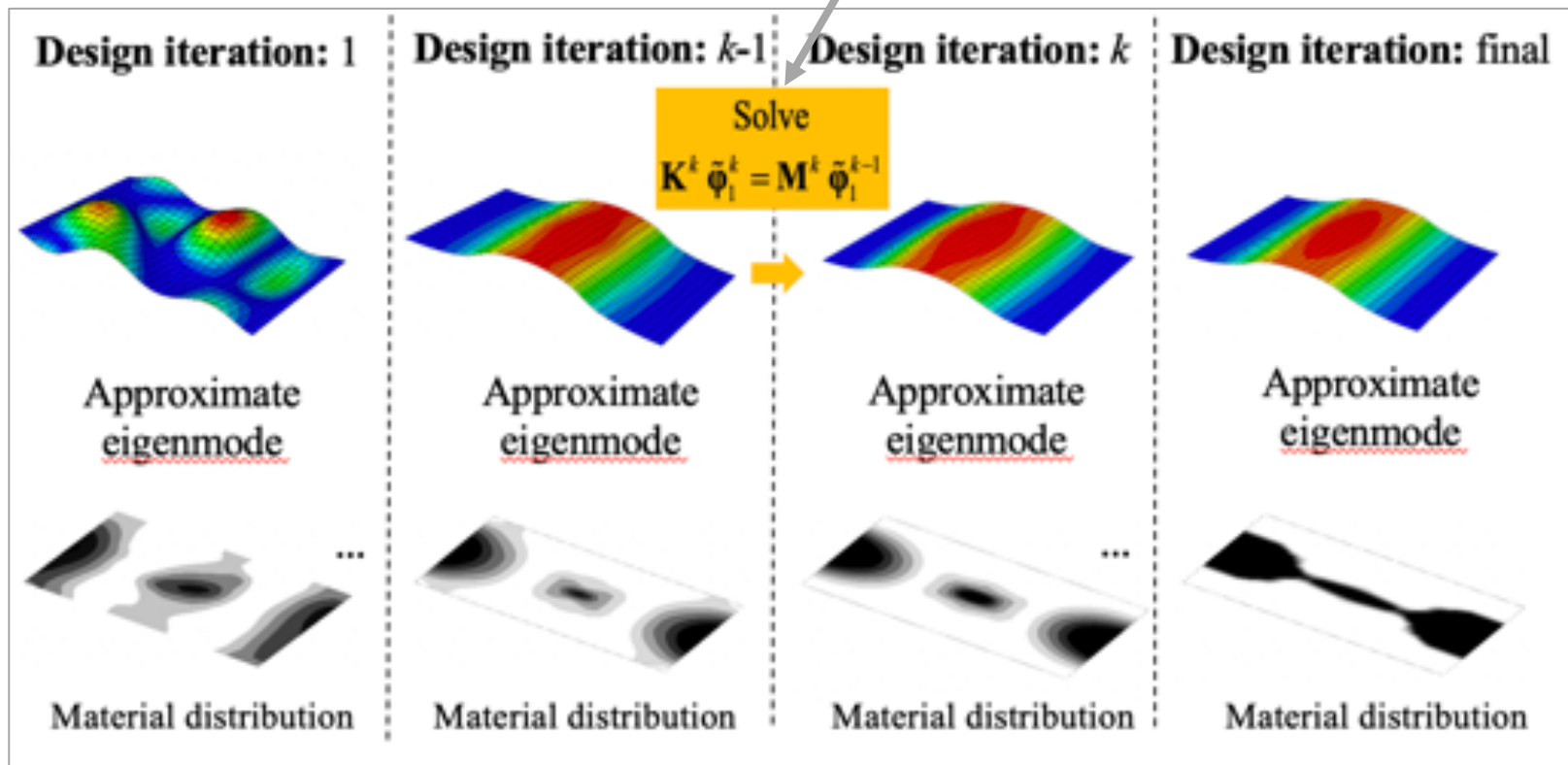
Eigenmodes computed with iterations



# Basic idea of SIAD method

Single loop iteration for eigenvalue analysis and design

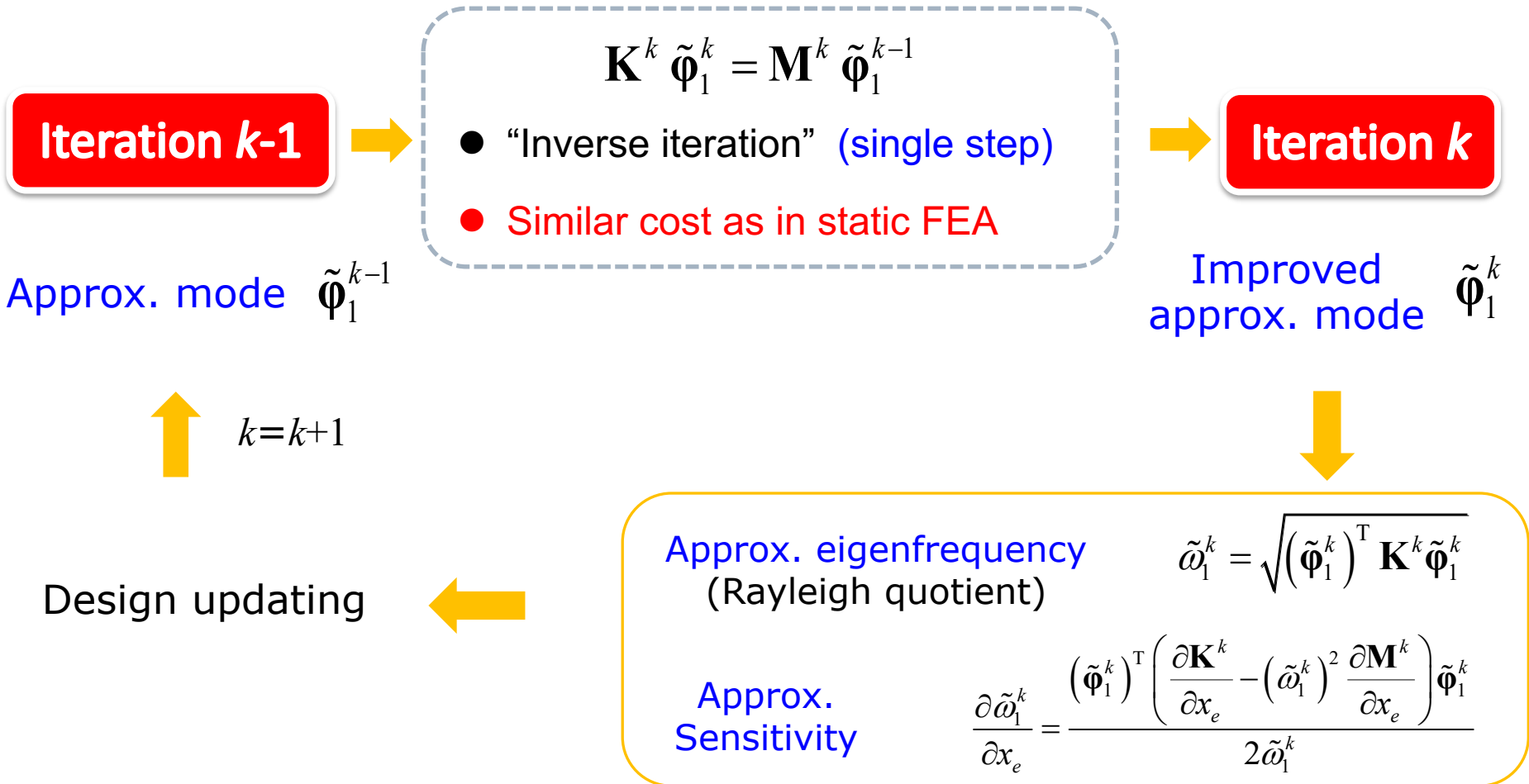
Approximate eigenmodes improved through a linear equation



Kang et al. *CMAME* 2020; 362: 112847

# Procedures of SIAD

## □ Case of fundamental frequency



# Discussion on convergence (case of distinct eigenvalue)

Exact eigenpairs for  $(k-1)$ th design:  $\lambda_1^{k-1}, \lambda_2^{k-1}, \dots, \lambda_n^{k-1}$        $\boldsymbol{\varphi}_1^{k-1}, \boldsymbol{\varphi}_2^{k-1}, \dots, \boldsymbol{\varphi}_n^{k-1}$

for  $k$ th design:       $\lambda_1^k, \lambda_2^k, \dots, \lambda_n^k$        $\boldsymbol{\varphi}_1^k, \boldsymbol{\varphi}_2^k, \dots, \boldsymbol{\varphi}_n^k$

$$\begin{cases} \Delta\lambda_j^k = (\boldsymbol{\varphi}_j^{k-1})^T (\Delta\mathbf{K}^k - \lambda_j^{k-1} \Delta\mathbf{M}^k) \boldsymbol{\varphi}_j^{k-1} \\ 2(\boldsymbol{\varphi}_j^{k-1})^T \mathbf{M}^{k-1} \Delta\boldsymbol{\varphi}_j^k = -(\boldsymbol{\varphi}_j^{k-1})^T \Delta\mathbf{M}^k \boldsymbol{\varphi}_j^{k-1} \end{cases}$$

Approximate eigenpairs for  $(k-1)$ th and  $k$ th designs:  $\tilde{\boldsymbol{\varphi}}^{k-1} = \sum_{j=1}^n a_j^k \boldsymbol{\varphi}_j^k$        $\tilde{\boldsymbol{\varphi}}^{k-1} = \sum_{j=1}^n a_j^{k-1} \boldsymbol{\varphi}_j^{k-1}$

Define:  $\eta_j^{k-1} = \frac{a_j^{k-1}}{a_1^{k-1}}, j = 1, 2, \dots, n$        $\eta_j^k = \frac{\lambda_1^k a_j^k}{\lambda_j^k a_1^k}, j = 1, 2, \dots, n$

We prove that convergence can be ensured on the condition that design change in each iteration is sufficiently small:

$$\eta_j^k \approx \frac{\lambda_1^{k-1} a_j^{k-1}}{\lambda_j^{k-1} a_1^{k-1}} < \eta_j^{k-1}, \quad j = 2, 3, \dots, n$$

CMAME 2020; 112847

# Case of higher-order frequencies

Subspace iteration-like procedures:

$$\mathbf{K}^k \tilde{\Psi}^k = \mathbf{M}^k \tilde{\Psi}^{k-1}$$

Inverse iteration (one step)  
of subspace

$$(\mathbf{K}^*)^k \tilde{\mathbf{A}}^k = (\mathbf{M}^*)^k \tilde{\mathbf{A}}^k \tilde{\mathbf{\Lambda}}^k$$

Eigenpairs in subspace  
(low-dimensional)

$$(\mathbf{M}^*)^k = (\tilde{\Psi}^k)^T \mathbf{M}^k \tilde{\Psi}^k$$

$$(\mathbf{K}^*)^k = (\tilde{\Psi}^k)^T \mathbf{K}^k \tilde{\Psi}^k$$

$$\tilde{\Psi}^k = \tilde{\Psi}^k \tilde{\mathbf{A}}^k$$

Mode updating

\* For repeated eigenvalues, a sub-problem of eigenvalue analysis needs to be solved to find Gateaux differentiable eigenmodes:

$$\mathbf{F} \mathbf{c}_i = \lambda_i' \mathbf{c}_i \quad \text{where } \mathbf{F} = (\tilde{\Phi}^k)^T \left( \partial \mathbf{K}^k / \partial x_e - (\tilde{\omega}_r^k)^2 \partial \mathbf{M}^k / \partial x_e \right) \tilde{\Phi}^k$$

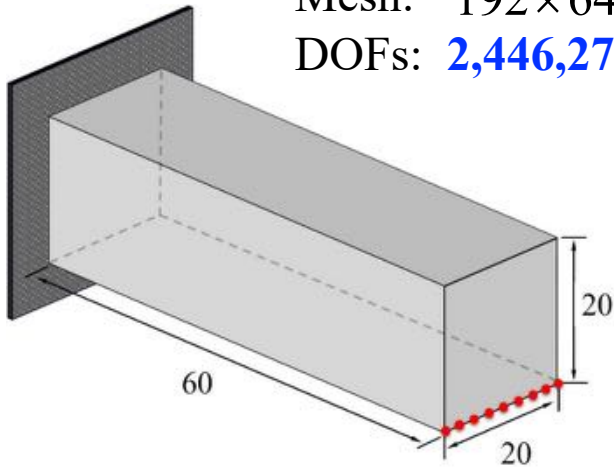
$$\tilde{\Phi}^k = \tilde{\Phi}^k \mathbf{C}$$



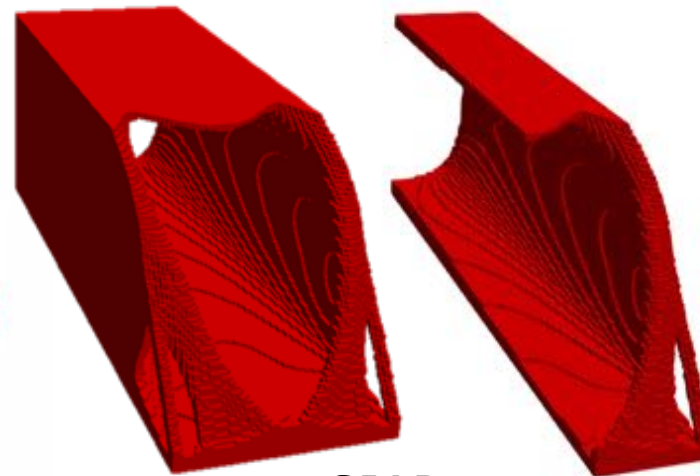
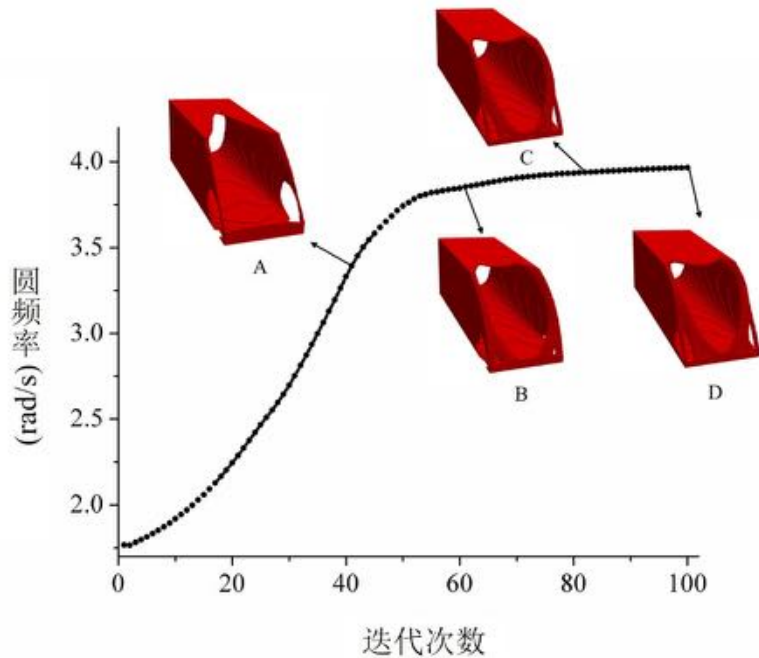
# Example: Fundamental frequency maximization

Mesh:  $192 \times 64 \times 64$

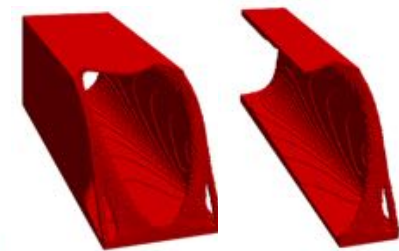
DOFs: **2,446,275**



	Number of full-scale equation solutions	Computing time (sec.)
SIAD	100	<b>22,980</b>
Double-loop	1058	<b>247,680</b>



SIAD  
3.97 rad/s



Double-loop  
4.08 rad/s

# Inensitive to initial trial eigenvectors

Random

Cosine

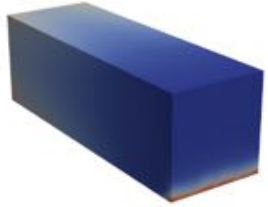
All-one

Random

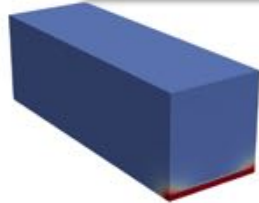
Cosine

All-one

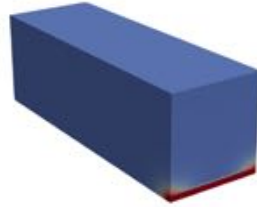
Iter 1



1.7676 rad/s

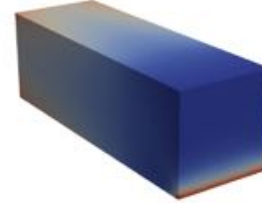


12.9885 rad/s

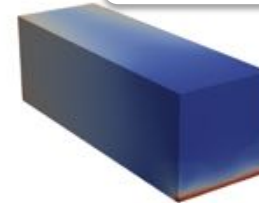


1.7734 rad/s

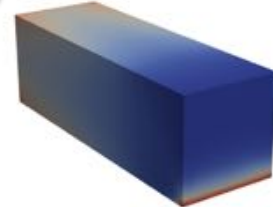
Iter 10



1.9723

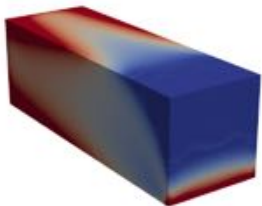


1.9274

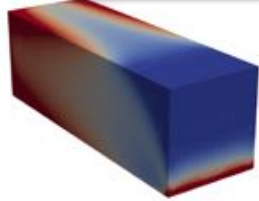


1.9729

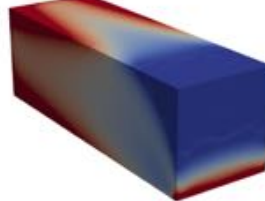
Iter 30



2.7582

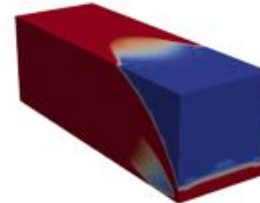


2.6342

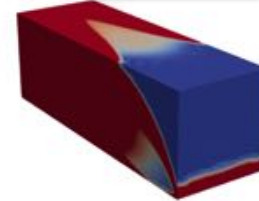


2.7578

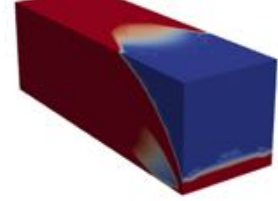
Iter 50



3.7638

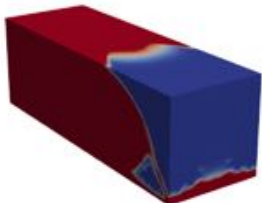


3.7406

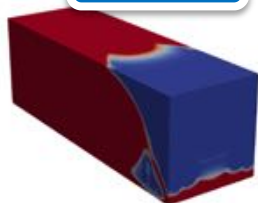


3.7639

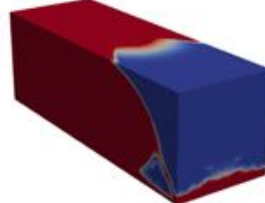
Iter 70



3.8515



3.8519



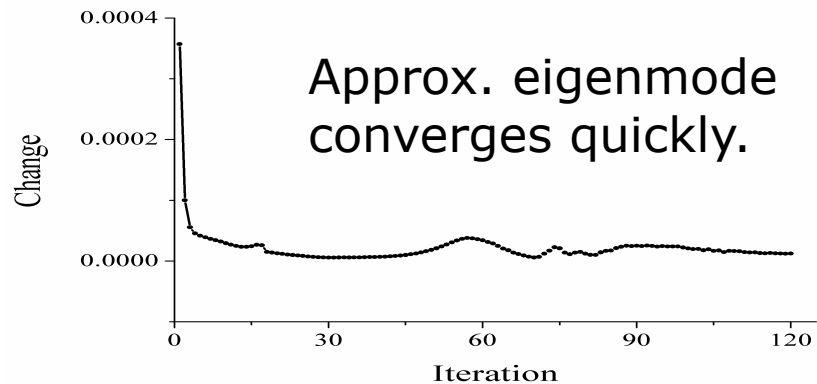
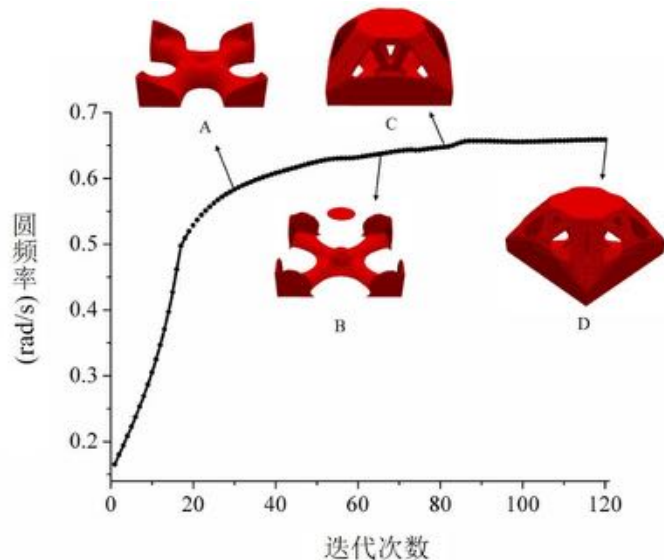
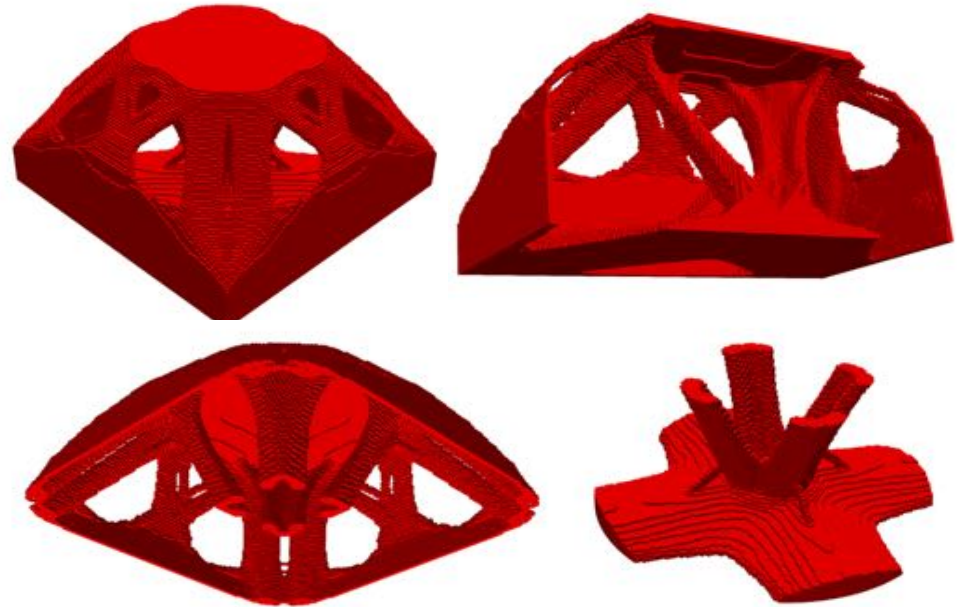
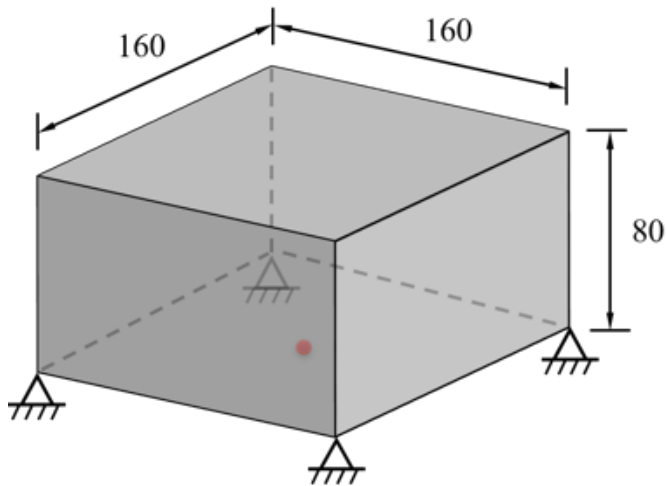
3.8513

# Example: Fundamental frequency maximization

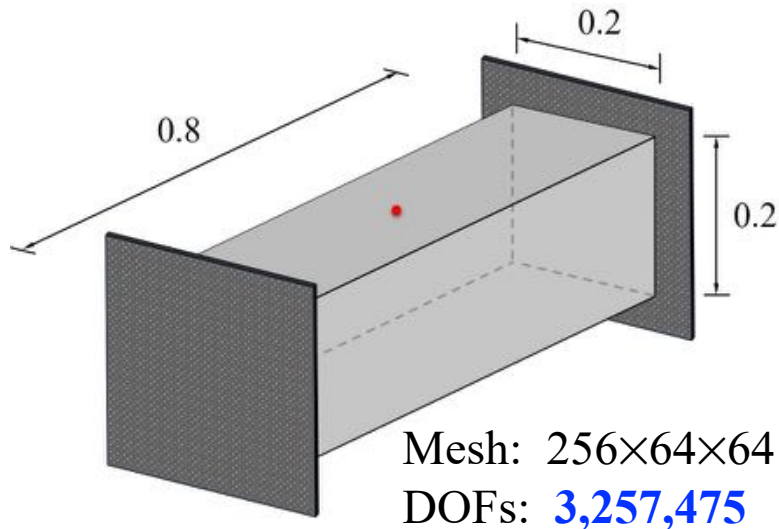
Mesh:  $160 \times 160 \times 80$

DOFs: **6,298,803**

Optimized design obtained with  
**random initial trial eigenmodes**

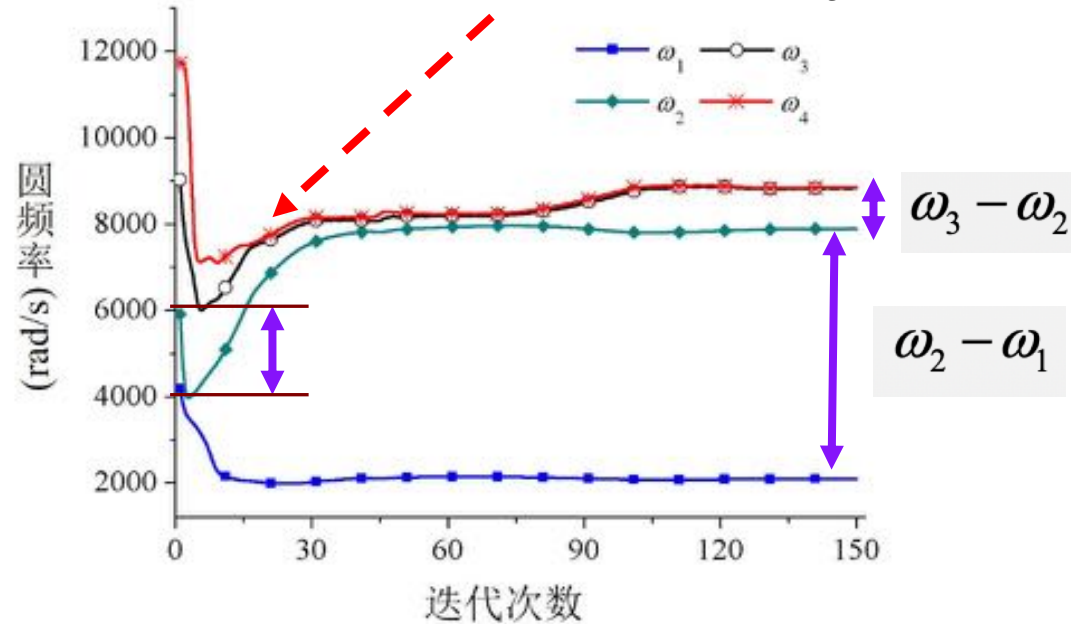


# Example: frequency gap maximization



$$\begin{aligned} \max : & \quad \omega_2 - \omega_1 \\ \text{s.t.} & \quad \omega_3 - \omega_2 \geq \beta \quad \beta = 1000 \\ & \quad \sum_{e=1}^{N_e} x_e V_e - f_v V_0 \leq 0 \\ & \quad 0 < \underline{x} \leq x_e \leq 1, \quad e = 1, \dots, N_e \end{aligned}$$

Repeated eigenvalues:  $\omega_3 = \omega_4$



# Conclusions

- SIAD implements **concurrent convergence** of eigenvalue analysis and design optimization.
- Affordable solution of eigenvalue optimization with **millions of DOFs** on a desktop computer.

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# THANK YOU !