

# PDE-Constrained Optimization Framework and Inverse Design of Mechanical Metamaterials for Vibration Control

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# Overview

- Motivation & Goals
- Designing structures & materials for vibration control
- New approach to design: Modified Error in Constitutive Equations
  - Transient domain formulation
  - Frequency domain formulation
- Numerical examples

# Motivation

- Harsh vibration and shock environments are common in engineering systems.
- We want to design materials and structures to control vibration and dissipate energy.
  - Steady-state behavior: elastic acoustic metamaterial design, base isolator design, etc
  - Transient behavior: shock absorption, impact response, etc.
- (Elastic) metamaterials are a potential solution!
- Application-specific designs may require complex geometries.

# How can we design these materials?

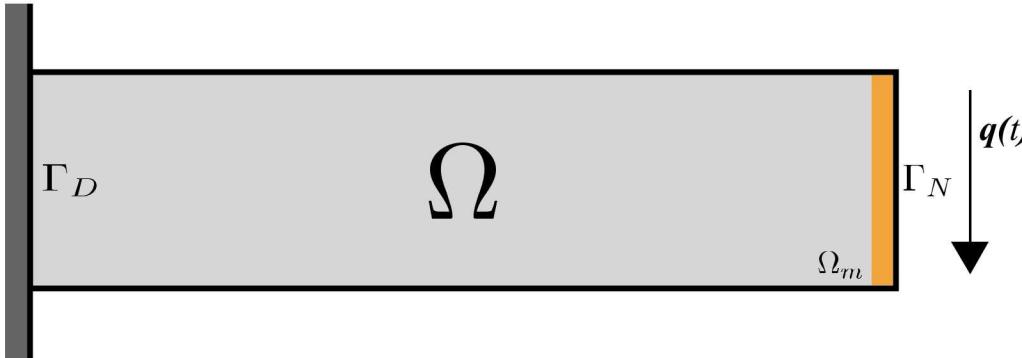
- Topology optimization: Determine the optimal distribution of material within a domain to satisfy structural objective
- Methods to design dynamic materials and structures:
  - Modal design: study frequency response spectra
  - Band-structure/Bloch Floquet analysis: design for material band gaps
  - Minimize dynamic compliance [1]
  - Direct design: design for specific frequency responses
- Solving the design problem: PDE-constrained optimization
  - Gradient-based methods

# Objectives

- Develop method to design structures and materials to match desired dynamic behavior
- Desire design method to be:
  - Robust with respect to initial guess, etc
  - Computationally efficient
- Utilize differentiability of constraints for gradient-based methods
- Leverage benefits of inverse-problem methods to avoid numerical and physical instabilities of direct methods

# Governing Equations for Transient Elastodynamics

Elastodynamics conservation of momentum equations govern system behavior and provide constraints for optimization. In transient case:



(1)  $\rho\omega^2\ddot{\mathbf{u}} = \nabla \cdot \boldsymbol{\sigma} + \mathbf{b}$  in  $\Omega \times [0, T]$  - Conservation of momentum

$\mathbf{u}(\cdot, 0) = \mathbf{0}$  on  $\Gamma_D \times [0, T]$  - Dirichlet boundary conditions

$\dot{\mathbf{u}}(\cdot, 0) = \mathbf{0}$  on  $\Gamma_D \times [0, T]$

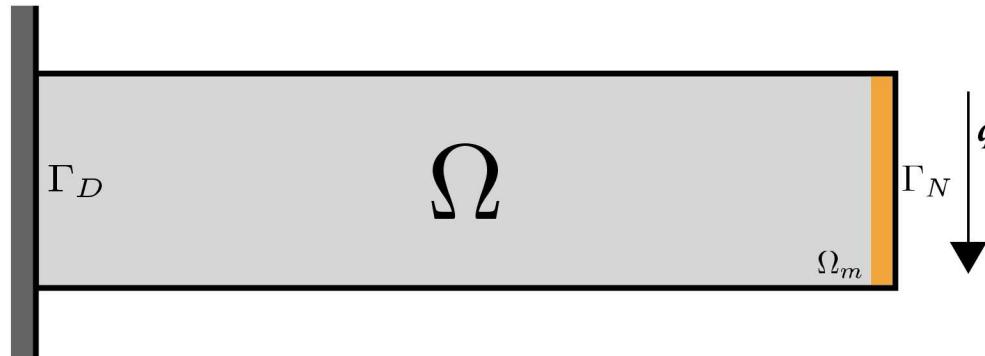
$\boldsymbol{\sigma} \cdot \mathbf{n} = \mathbf{q}(t)$  on  $\Gamma_N \times [0, T]$  - Neumann boundary condition

$\boldsymbol{\sigma} = \mathbb{C} : \epsilon[\mathbf{u}]$  - Constitutive equations

$\epsilon[\mathbf{u}] = \frac{1}{2} (\nabla \mathbf{u} + \nabla \mathbf{u}^T)$

# Governing Equations for Frequency-Domain Elastodynamics

In steady-state dynamics, the equations of motion may be expressed:



$$(1) \quad -\nabla \cdot \boldsymbol{\sigma} - \rho\omega^2 \mathbf{u} = \mathbf{b} \text{ in } \Omega$$

$$\mathbf{u} = \mathbf{0} \text{ on } \Gamma_D$$

$$\boldsymbol{\sigma} \cdot \mathbf{n} = \mathbf{q} \text{ on } \Gamma_N$$

$$\boldsymbol{\sigma} = \mathbb{C} : \boldsymbol{\epsilon}[\mathbf{u}]$$

$$\boldsymbol{\epsilon}[\mathbf{u}] = \frac{1}{2} (\nabla \mathbf{u} + \nabla \mathbf{u}^T)$$

- Conservation of momentum
- Dirichlet boundary conditions
- Neumann boundary condition
- Constitutive equations

# Material Modeling and Design Variables

- Linear elastic materials, with strictly enforced constitutive equation, parameterized by bulk and shear modulus:

$$\boldsymbol{\sigma} = \mathbb{C} : \boldsymbol{\epsilon}[\mathbf{u}] = (G\mathbf{D}_G + b\mathbf{D}_b) : \boldsymbol{\epsilon}[\mathbf{u}] \quad (2)$$

- Solid Isotropic Material Penalization (SIMP) scheme for multi-phase material interpolation [2]
  - Penalization parameter  $p \geq 1$  discourages intermediate values.
  - Material phases characterized by set of bulk modulus, shear modulus, and mass density:

$$G(\boldsymbol{\beta}; p) := G^0 + (G^1 - G^0)\beta^p$$

$$b(\boldsymbol{\beta}; p) := b^0 + (b^1 - b^0)\beta^p$$

$$\rho(\boldsymbol{\beta}; p) := \rho^0 + (\rho^1 - \rho^0)\beta^p$$

$$\beta_i \in [0, 1], i = 1..d$$

$$p \in \mathbb{Z}^+$$

# Least-squares Design Problem

- Define the minimization problem: Find displacements  $\mathbf{u}$  and design  $\boldsymbol{\beta}$  which minimize the least-squares misfit of the target displacement response

$$\arg \min_{\mathbf{u}, \boldsymbol{\beta}} \mathcal{J}(\mathbf{u}, \boldsymbol{\beta})$$

subject to: Eqn. [1] & Eqn. [2] **(L2)**

$$\beta_i \leq 1, i = 1 \dots d$$

$$\beta_i \geq 0, i = 1 \dots d$$

# Can we solve this problem robustly?

- Least-squares problem can be unstable.
- In frequency domain, resonances different designs' responses within searchable design space cause non-convexity in objective function
  - Inferior local minima may inhibit progress of gradient-based methods
  - Ideally, we want to design for multiple frequencies (difficult, given this phenomenon)

# A different approach...

We can allow violation of the constitutive equations, which improves the convexity of the objective function.

**Modified Error in Constitutive Equations (MECE):** penalty method which weights competing objectives of matching response exactly and explicitly enforcing constitutive equations.

$$\Lambda(\mathbf{u}, \boldsymbol{\sigma}, \boldsymbol{\beta}; \kappa) = \mathcal{E}(\mathbf{u}, \boldsymbol{\sigma}, \boldsymbol{\beta}) + \frac{\kappa}{2} \mathcal{J}(\mathbf{u}, \boldsymbol{\beta})$$

Penalty parameter  $\kappa$  weights the two objectives.

**MECE** has been successfully utilized in **material identification inverse problems** in frequency and transient scenarios [3,4].

# MECE in Transient Problems

In a transient problem, we strive to design a structure such that its dynamic response matches target displacement time histories.

The objective function components are expressed:

- 1) **Error in constitutive equations term:** measures violation of constitutive equation relating stresses and strains:

$$\mathcal{E}(\mathbf{u}, \boldsymbol{\sigma}, \boldsymbol{\beta}) = \frac{1}{2} \int_0^T ((\boldsymbol{\sigma}_t - \mathbb{C}(\boldsymbol{\beta}) : \boldsymbol{\epsilon}[\mathbf{u}_t]), \mathbb{C}(\boldsymbol{\beta})^{-1} : (\boldsymbol{\sigma}_t - \mathbb{C}(\boldsymbol{\beta}) : \boldsymbol{\epsilon}[\mathbf{u}_t])) dt$$

- 2) **Least-squares term:** misfit of calculated displacements and target displacement patterns, weighted by penalty parameter

$$\mathcal{J}(\mathbf{u}, \boldsymbol{\beta}) = \frac{\kappa}{2} \|\mathbf{u} - \mathbf{u}^m\|_{L_2(\Omega_m)}^2$$

# MECE in Steady-State Problems

We seek to design a structure such that it can match a target vibrational response.

The objective function components are expressed:

**1) Error in constitutive equations term:**

$$\mathcal{E}(\mathbf{u}, \boldsymbol{\sigma}, \boldsymbol{\beta}) = \frac{1}{2} \left( (\boldsymbol{\sigma} - \mathbb{C} : \boldsymbol{\epsilon}[\mathbf{u}]), \mathbb{C}^{-1} : (\boldsymbol{\sigma} - \mathbb{C} : \boldsymbol{\epsilon}[\mathbf{u}]) \right)_{\Omega}$$

**2) Least-squares term:** misfit of calculated displacements and target displacement patterns

$$\mathcal{J}(\mathbf{u}, \boldsymbol{\beta}) = \frac{\kappa}{2} \|\mathbf{u} - \mathbf{u}^m\|_{L_2(\Omega_m)}^2$$

Optimization for multiple frequencies is represented as sum of objective functions corresponding to individual frequency responses.

# MECE Design Problem

- Define the minimization problem:

$$\arg \min_{\boldsymbol{u} \in \mathcal{V}, \boldsymbol{\sigma} \in \mathcal{S}, \boldsymbol{\beta}} \Lambda(\boldsymbol{u}, \boldsymbol{\sigma}, \boldsymbol{\beta}; \kappa)$$

subject to: Eqn. [1]

**(MECE)**

$$\beta_i \leq 1, i = 1 \dots d$$

$$\beta_i \geq 0, i = 1 \dots d$$

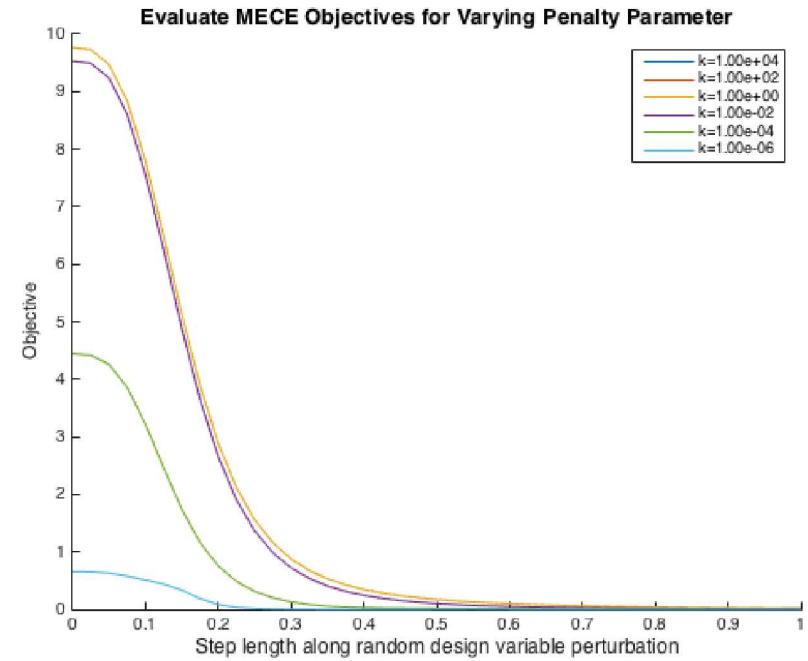
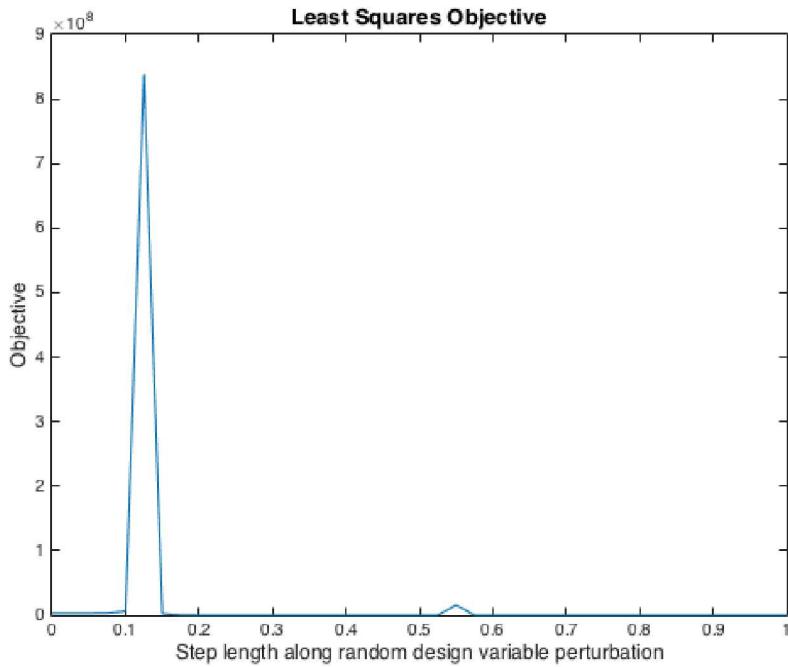
- Constitutive equations now not explicitly enforced as equality constraints

# A different approach...

We can compare the objective functions of the L2 and MECE problems.

**Objective:** minimize the displacement at the end of a vibrating beam.

MECE, with decreasing penalty parameter, converges to Least-squares:



# MECE Problem Solution Methods

- Gradient based optimization method
- Form Lagrangian, incorporating objective and PDE-constraints

$$\mathcal{L}(\mathbf{u}, \boldsymbol{\sigma}, \boldsymbol{\beta}, \mathbf{w}; \kappa) := \Lambda(\mathbf{u}, \boldsymbol{\sigma}, \boldsymbol{\beta}; \kappa) + ((\boldsymbol{\sigma}, \boldsymbol{\epsilon}[\mathbf{w}]) - \rho(\boldsymbol{\beta})\omega^2(\mathbf{u}, \mathbf{w}) - \mathbf{f}(\mathbf{w}))$$

- KKT first-order optimality conditions lead to:
  - Expression relating stresses to displacement fields  $\{\mathbf{u}, \mathbf{w}\}$ :

$$\mathcal{L}'_{\sigma}[\hat{\boldsymbol{\sigma}}] = 0 \rightarrow \boldsymbol{\sigma} = \mathbb{C} : \boldsymbol{\epsilon} [\mathbf{u} - \mathbf{w}]$$

- Coupled linear system for forward and adjoint solutions (state variable and Lagrange multipliers)

$$\begin{Bmatrix} \mathcal{L}'_u[\tilde{\mathbf{u}}] \\ \mathcal{L}'_w[\tilde{\mathbf{w}}] \end{Bmatrix} = \begin{Bmatrix} \mathbf{0} \\ \mathbf{0} \end{Bmatrix} \rightarrow \begin{bmatrix} [\mathbf{H}](\boldsymbol{\beta}) & \kappa[\mathbf{Q}] \\ -[\mathbf{K}](\boldsymbol{\beta}) & [\mathbf{H}](\boldsymbol{\beta}) \end{bmatrix} \begin{Bmatrix} \hat{\mathbf{w}}_{\beta} \\ \hat{\mathbf{u}}_{\beta} \end{Bmatrix} = \begin{Bmatrix} [\kappa\mathbf{Q}]\hat{\mathbf{u}}^t \\ \mathbf{F} \end{Bmatrix}$$

# MECE in Design

In inverse problem approaches, the penalty parameter is selected from criteria based on measurement error (e.g. Morozov discrepancy principle)[3]

How can we interpret the penalty parameter in design?

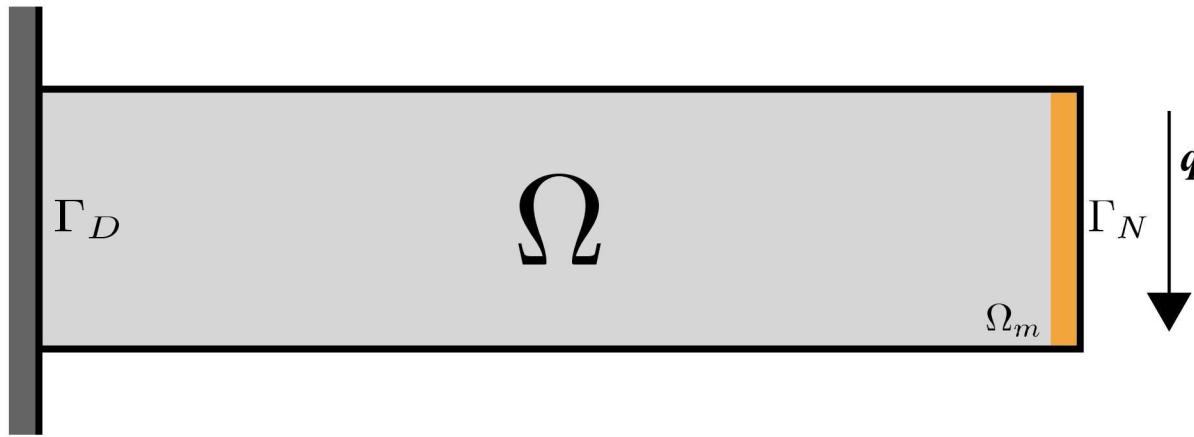
1. **Optimal Penalty Parameter:** Select an optimal  $\kappa$  parameter to design, entailing possible violation of constitutive laws
2. **Continuation Approach:** Solve a sequence of optimization problems, with decreasing  $\kappa$ , to converge to original least-squares problem solution

We'll focus on the second approach.

# Design Example: Cantilever Beam Design

## Objective:

Minimize vibrational response at end of cantilever beam experiencing forced vibration at multiple frequencies.



## Parameters:

Material 0:  $\{G_0, b_0, \rho_0\} = \{1, 2, 1e-4\}$

Material 1:  $\{G_1, b_1, \rho_0\} = \{1e-2, 2e-2, 1e-6\}$

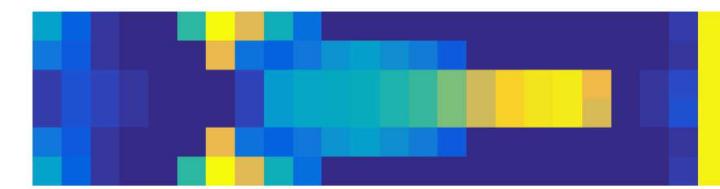
Initial Guess:  $\beta = 1$

Loading Frequencies:

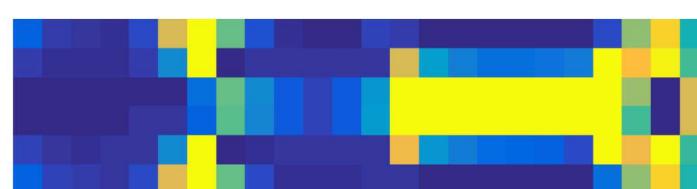
Case 1: Single Frequency, 5 Hz

Case 2: Multiple Frequencies,  $\{2, 4, 6\}$  Hz

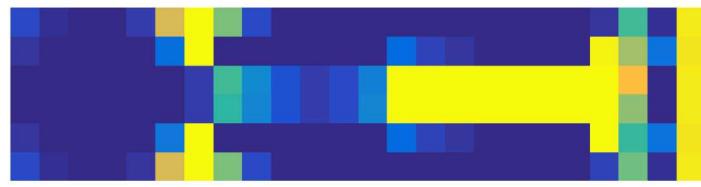
# Single-Frequency Optimization: Design Iterations



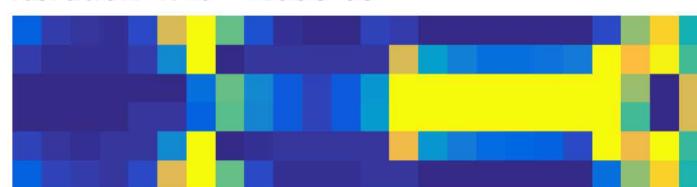
Iteration 1:  $\kappa = 1.00\text{e+}03$



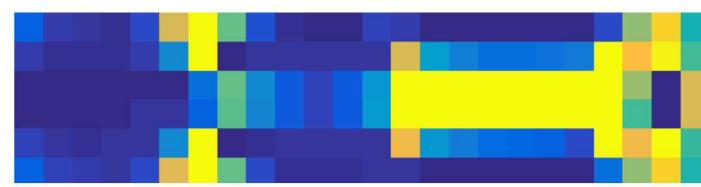
Iteration 4:  $\kappa = 1.00\text{e-}03$



Iteration 2:  $\kappa = 1.00\text{e+}01$



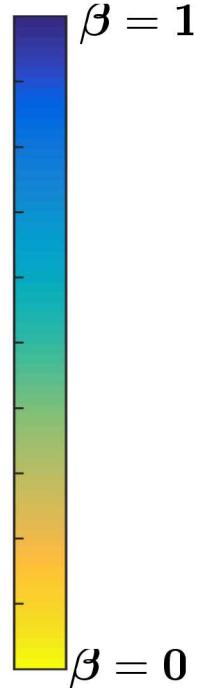
Iteration 5:  $\kappa = 1.00\text{e-}05$



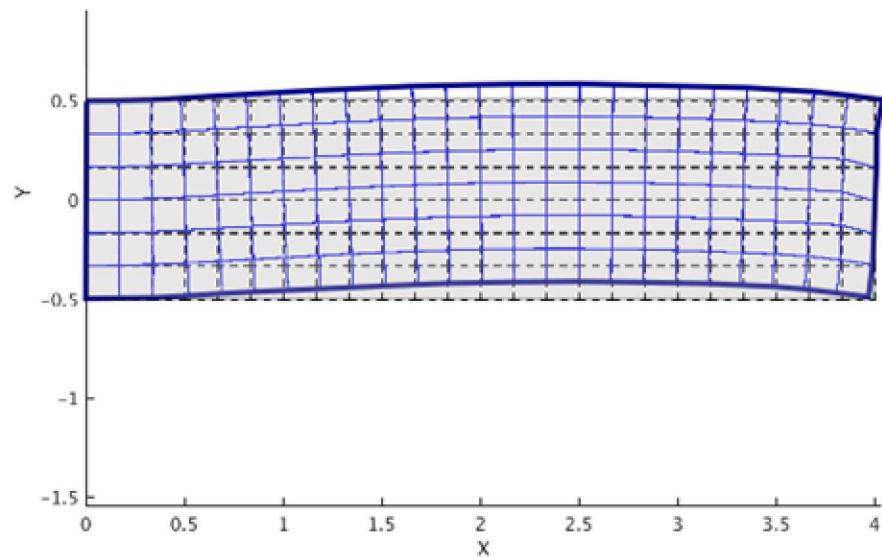
Iteration 3:  $\kappa = 1.00\text{e-}01$



L2 Solution

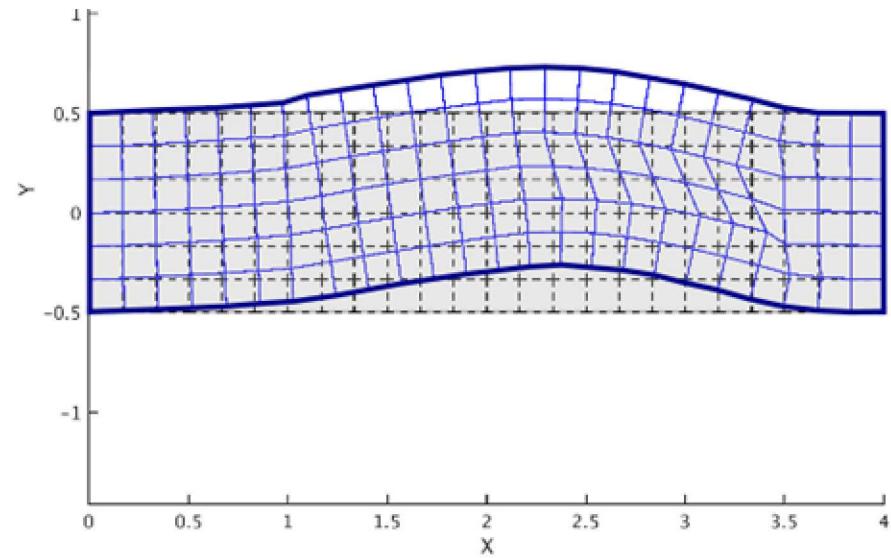


# Single Frequency Design: Displacement Patterns



**FIGURE:** Displacement pattern for L2-problem design  
Relative displacement magnitude reduction:

$$\frac{\mathbf{u}_{L2}^T[\mathbf{Q}]\mathbf{u}_{L2}}{\mathbf{u}_0^T[\mathbf{Q}]\mathbf{u}_0} = 0.0906$$



**FIGURE:** Displacement pattern for C-MECE problem design.  
Relative displacement magnitude reduction:

$$\frac{\mathbf{u}_M^T[\mathbf{Q}]\mathbf{u}_M}{\mathbf{u}_0^T[\mathbf{Q}]\mathbf{u}_0} = 3.2660\text{e-}05$$

# Multi-Frequency Optimization: Design Iterations



Iteration 1:  $\kappa = 1.00e+04$



Iteration 2:  $\kappa = 1.00e+02$



Iteration 3:  $\kappa = 1.00e+00$



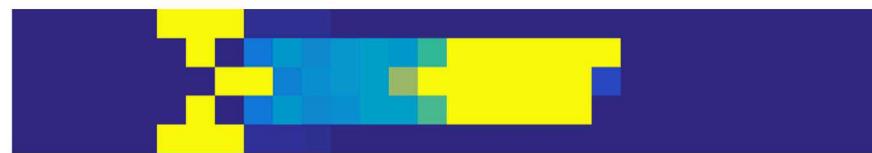
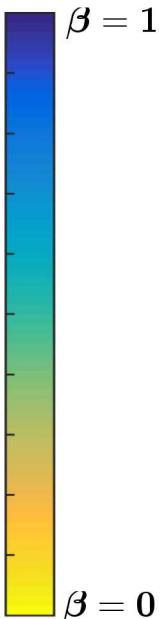
Iteration 4:  $\kappa = 1.00e-02$



Iteration 5:  $\kappa = 1.00e-02$



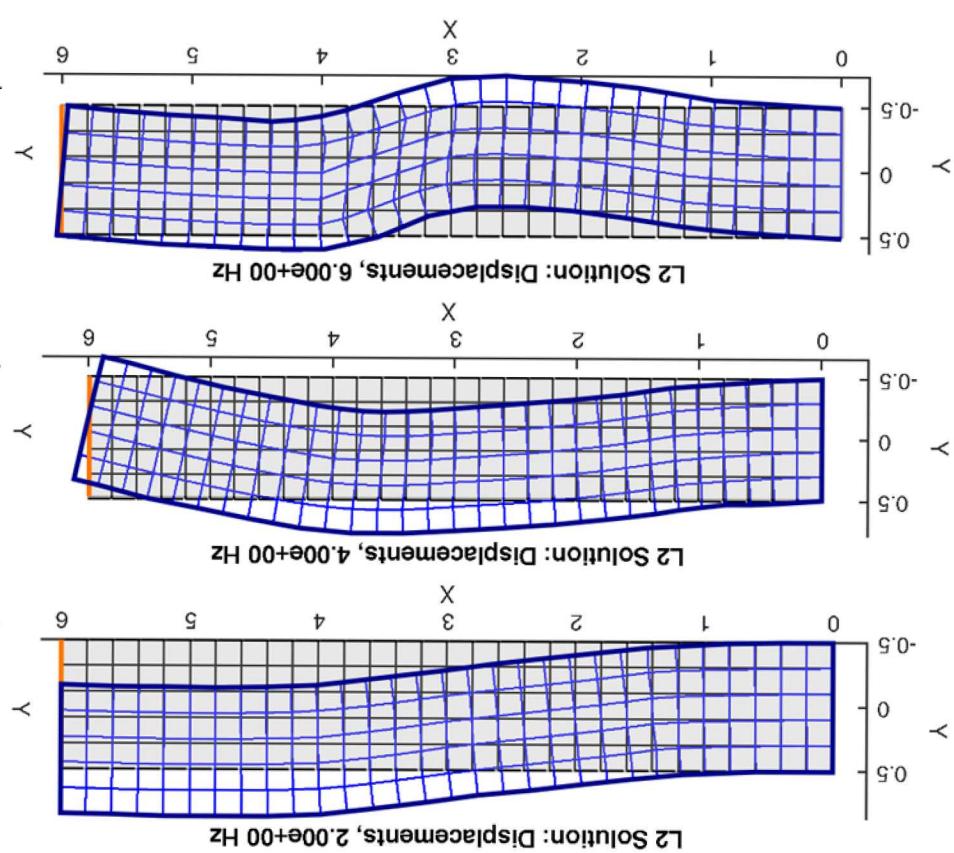
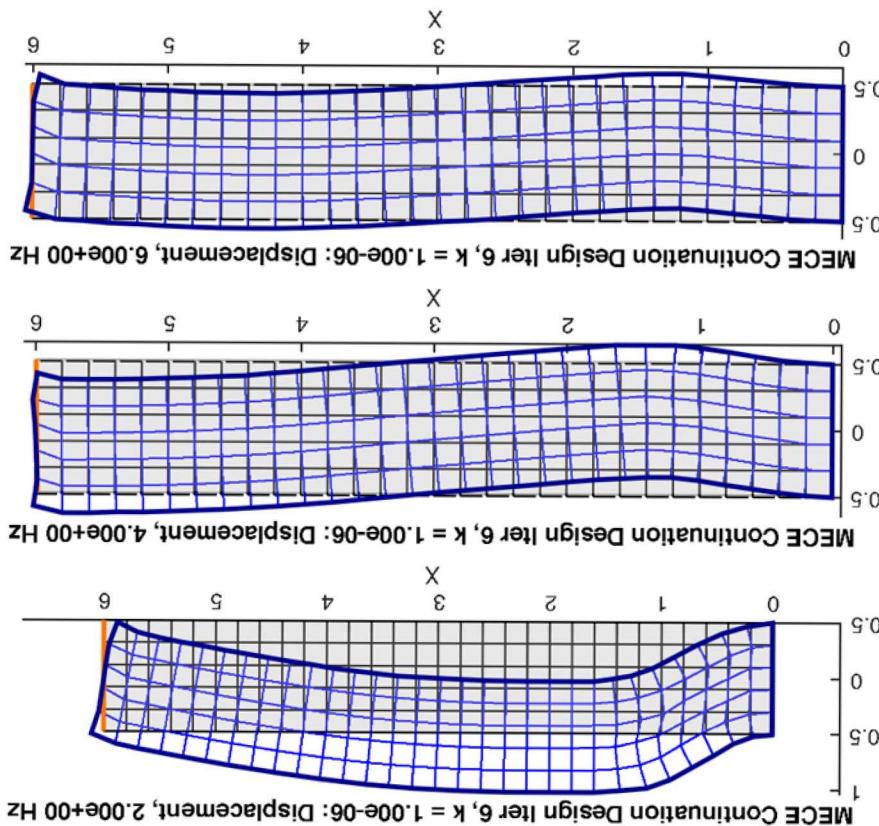
Iteration 6:  $\kappa = 1.00e-02$



L2 Solution

# Multi-Frequency Optimization: Displacement

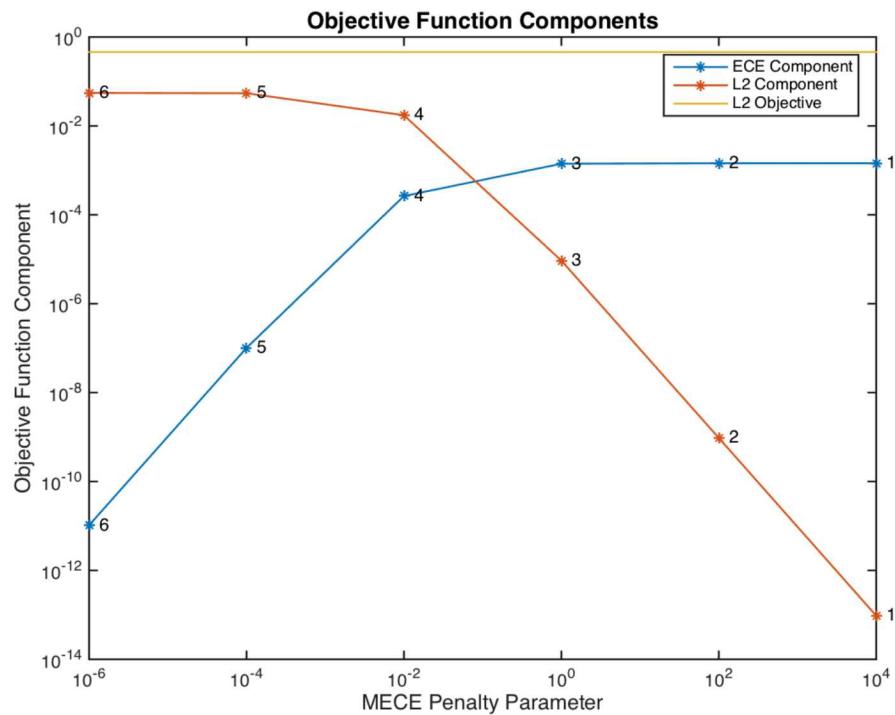
Patterns



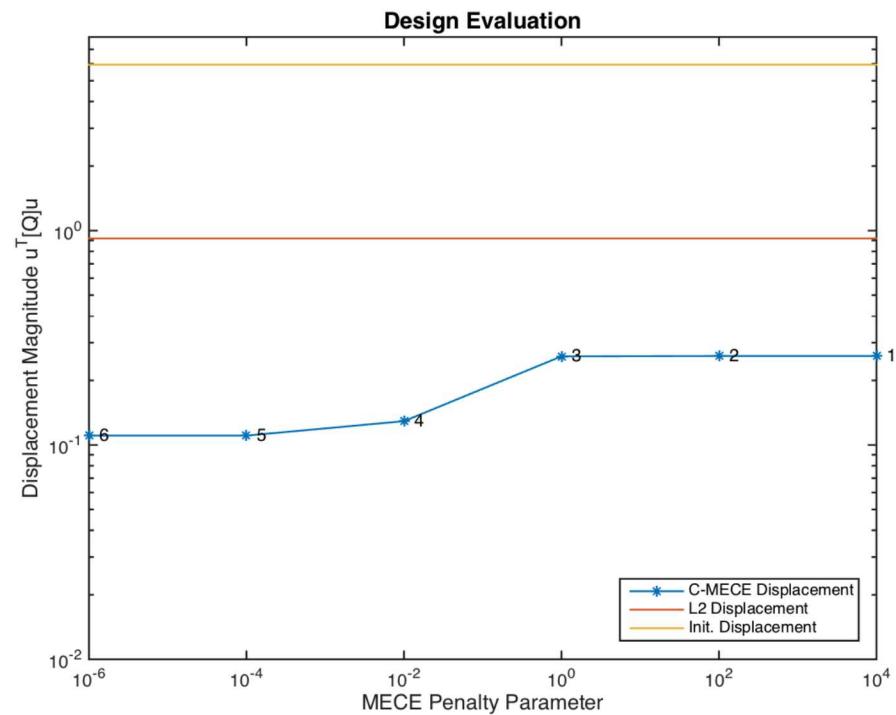
**FIGURE:** Displacement patterns for L2-problem design  
Relative end displacement magnitude: 0.1545

**FIGURE:** Displacement patterns for C-MCE problem design  
Relative end displacement magnitude: 0.0186

# Evolution of Design Performance in Continuation Method



**FIGURE:** Evolution of objective function components for successive continuation iterations



**FIGURE:** Design performance, measured with displacement magnitude, for successive continuation iterations.

# Conclusions

- Presented a Modified Error in Constitutive Equations approach to material and structural design
- Continuation scheme solves sequence of MECE optimization problems, reducing the penalty parameter and converging to L2 problem
- Continuation scheme approach can converge to a unique local minimum from L2 problem
- Multi-frequency problems especially promising for MECE; superposition of FRF's in L2 problem makes problem less convex
- Future directions:
  - Extension to acoustic-structural interaction for acoustic cloaking applications
  - Transient problem implementation

# References

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