

# Rapid Topology Optimization using Reduced-Order Models

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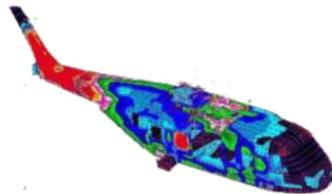
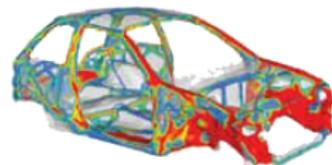
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# Motivation

- For industry-scale design problems, topology optimization is a beneficial tool that is *time and resource intensive*
  - Large number of calls to structural solver usually required
  - Each structural call is expensive, especially for nonlinear 3D High-Dimensional Models (HDM)
- Use a Reduced-Order Model (ROM) as a surrogate for the structural model in a material topology optimization loop
  - Large speedups over HDM realized



## 0-1 Material Topology Optimization

$$\begin{aligned} & \underset{\boldsymbol{\chi} \in \mathbb{R}^{n_{el}}}{\text{minimize}} && \mathcal{L}(\mathbf{u}(\boldsymbol{\chi}), \boldsymbol{\chi}) \\ & \text{subject to} && \mathbf{c}(\mathbf{u}(\boldsymbol{\chi}), \boldsymbol{\chi}) \leq 0 \end{aligned}$$

- $\mathbf{u}$  is implicitly defined as a function of  $\boldsymbol{\chi}$  through the HDM equation

$$\mathbf{f}^{int}(\mathbf{u}) = \mathbf{f}^{ext}$$

$$\mathbb{C}^e = \mathbb{C}_0^e \boldsymbol{\chi}_e \quad \rho^e = \rho_0^e \boldsymbol{\chi}_e \quad \boldsymbol{\chi}_e = \begin{cases} 0, & e \notin \Omega^* \\ 1, & e \in \Omega^* \end{cases}$$

- Assume geometric *nonlinearity* and linear material law
  - Large deformations of St. Venant-Kirchhoff material



## Reduced-Order Model

- Model Order Reduction (MOR) assumption
  - State vector lies in low-dimensional subspace defined by a Reduced-Order Basis (ROB)  $\Phi \in \mathbb{R}^{N \times k_u}$

$$\mathbf{u} \approx \Phi \mathbf{y}$$

- $k_u \ll N$
- $N$  equations,  $k_u$  unknowns

$$\mathbf{f}^{int}(\Phi \mathbf{y}) = \mathbf{f}^{ext}$$

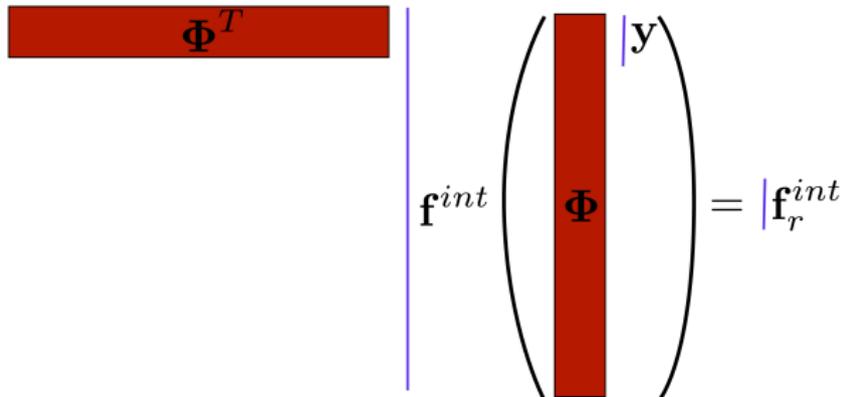
- Galerkin projection

$$\Phi^T \mathbf{f}^{int}(\Phi \mathbf{y}) = \Phi^T \mathbf{f}^{ext}$$



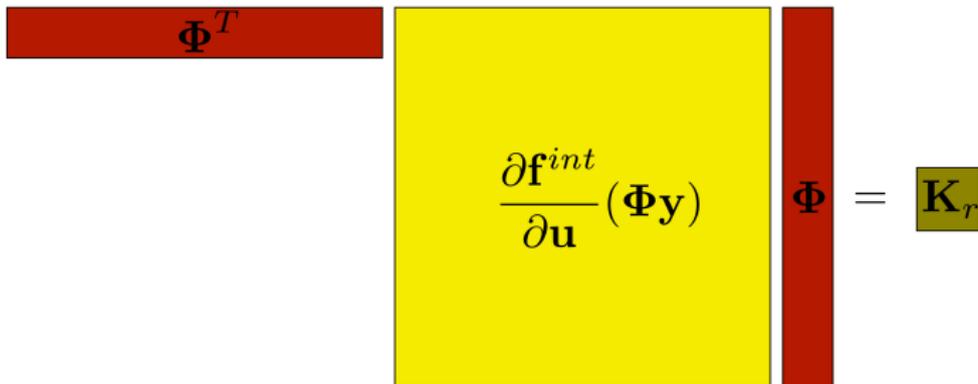
# NL ROM Bottleneck - Internal Force

$$\Phi^T \mathbf{f}^{int}(\Phi \mathbf{y}) = \Phi^T \mathbf{f}^{ext}$$



# NL ROM Bottleneck - Tangent Stiffness

$$\Phi^T \mathbf{f}^{int}(\Phi \mathbf{y}) = \Phi^T \mathbf{f}^{ext}$$



## Internal Force

The expression for the internal force is

$$\mathbf{f}_{jL}^{int} = \int_{\Omega_0} \mathbf{P}_{ij} \frac{\partial \mathbf{N}_L}{\partial \mathbf{X}_i} d\mathbf{X}$$

where  $\mathbf{N}_I(\mathbf{X})$  is the shape function corresponding to node  $I$  and

$$\mathbf{u}_i(\mathbf{X}) = \mathbf{u}_{iI} \mathbf{N}_I(\mathbf{X}) \quad (\text{FEM discretization})$$

$$\mathbf{F} = \mathbf{I} + \mathbf{u} \frac{\partial \mathbf{N}}{\partial \mathbf{X}} \quad (\text{Deformation Gradient})$$

$$\mathbf{E} = \frac{1}{2} (\mathbf{F}^T \mathbf{F} - \mathbf{I}) \quad (\text{Green-Lagrange Strain})$$

$$\mathbf{P} = \mathbf{S} \mathbf{F}^T \quad (\text{First Piola-Kirchhoff Stress})$$

$$\mathbf{S} = \lambda(\mathbf{X}) \text{tr}(\mathbf{E}) \mathbf{I} + 2\mu(\mathbf{X}) \mathbf{E} \quad (\text{Second Piola-Kirchhoff Stress})$$



## Internal Force - Cubic Polynomial in Displacements

$$\begin{aligned}
 \mathbf{f}_{jL}^{int} &= \int_{\Omega_0} \mathbf{P}_{ij} \frac{\partial \mathbf{N}_I}{\partial \mathbf{X}_i} d\mathbf{X} \\
 &= \bar{\mathbf{A}}_{jtIL} \mathbf{u}_{tI} + \bar{\mathbf{B}}_{LI} \mathbf{u}_{jI} + \\
 &\quad \bar{\mathbf{C}}_{LIJj} \mathbf{u}_{kI} \mathbf{u}_{kJ} + \hat{\mathbf{C}}_{ILQt} \mathbf{u}_{jQ} \mathbf{u}_{tI} + \bar{\mathbf{D}}_{IJQL} \mathbf{u}_{kI} \mathbf{u}_{kJ} \mathbf{u}_{jQ}
 \end{aligned}$$

where,

$$\bar{\mathbf{A}} = \bar{\mathbf{A}}(\Omega, \lambda(\mathbf{X}))$$

$$\bar{\mathbf{B}} = \bar{\mathbf{B}}(\Omega, \mu(\mathbf{X}))$$

$$\bar{\mathbf{C}} = \bar{\mathbf{C}}(\Omega, \lambda(\mathbf{X}), \mu(\mathbf{X}))$$

$$\hat{\mathbf{C}} = \hat{\mathbf{C}}(\Omega, \lambda(\mathbf{X}), \mu(\mathbf{X}))$$

$$\bar{\mathbf{D}} = \bar{\mathbf{D}}(\Omega, \lambda(\mathbf{X}), \mu(\mathbf{X}))$$



## Material Representation

Let material distributions be represented with the basis functions:

$$\lambda(\mathbf{X}) = \phi_i^\lambda(\mathbf{X})\alpha_i^r, \quad i = 1, 2, \dots, n_\alpha$$

$$\mu(\mathbf{X}) = \phi_i^\mu(\mathbf{X})\alpha_i^r, \quad i = 1, 2, \dots, n_\alpha$$

$$\rho(\mathbf{X}) = \phi_i^\rho(\mathbf{X})\alpha_i^r, \quad i = 1, 2, \dots, n_\alpha.$$

Then

$$\bar{\mathbf{A}} = \bar{\mathbf{A}}(\Omega, \phi_i^\lambda) \alpha_i^r$$

$$\bar{\mathbf{B}} = \bar{\mathbf{B}}(\Omega, \phi_i^\mu) \alpha_i^r$$

$$\bar{\mathbf{C}} = \bar{\mathbf{C}}(\Omega, \phi_i^\lambda, \phi_i^\mu) \alpha_i^r$$

$$\hat{\mathbf{C}} = \hat{\mathbf{C}}(\Omega, \phi_i^\lambda, \phi_i^\mu) \alpha_i^r$$

$$\bar{\mathbf{D}} = \bar{\mathbf{D}}(\Omega, \phi_i^\lambda, \phi_i^\mu) \alpha_i^r$$



## Pre-computed ROM - cubic nonlinearity

- HDM

$$\mathbf{f}_{jL}^{int} = \bar{\mathbf{A}}_{jtIL} \mathbf{u}_{tI} + \bar{\mathbf{B}}_{LI} \mathbf{u}_{jI} + \bar{\mathbf{C}}_{LIJj} \mathbf{u}_{kI} \mathbf{u}_{kJ} \\ + \hat{\mathbf{C}}_{ILQt} \mathbf{u}_{jQ} \mathbf{u}_{tI} + \bar{\mathbf{D}}_{IJQL} \mathbf{u}_{kI} \mathbf{u}_{kJ} \mathbf{u}_{jQ}$$

- ROM

$$[\Phi^T \mathbf{f}^{int}(\Phi \mathbf{y})]_r = \beta_{rp} \mathbf{y}_p + \gamma_{rpq} \mathbf{y}_p \mathbf{y}_q + \omega_{rpqt} \mathbf{y}_p \mathbf{y}_q \mathbf{y}_t$$

$$\beta = \beta(\Phi, \phi_i^\lambda, \phi_i^\mu) \alpha_i^r$$

$$\gamma = \gamma(\Phi, \phi_i^\lambda, \phi_i^\mu) \alpha_i^r$$

$$\omega = \omega(\Phi, \phi_i^\lambda, \phi_i^\mu) \alpha_i^r$$

$$\Phi^T \mathbf{f}^{int}(\Phi \mathbf{y}) = \Phi^T \mathbf{f}^{ext}$$



# ROM Pre-computation Approach

## Advantages

- Only need to solve small, cubic nonlinear system online
- Large speedups possible without hyperreduction,  $\mathcal{O}(10^3)$
- Amenable to 0-1 material topology optimization
  - $\alpha^r$  provide control over material distribution
  - $\alpha^r$  can be used as optimization variables

## Disadvantages

- *Currently* limited to StVK material, *Lagrangian* elements
- Offline cost scales as  $\mathcal{O}(n_\alpha \cdot n_{el} \cdot k_{\mathbf{u}}^4)$
- Offline storage scales as  $\mathcal{O}(n_\alpha \cdot k_{\mathbf{u}}^4)$
- Online storage scales as  $\mathcal{O}(k_{\mathbf{u}}^4)$
- Can only vary material distribution in the subspace defined by the material snapshot vectors



# Reduced Topology Optimization

$$\begin{aligned} & \underset{\boldsymbol{\alpha}_r \in \mathbb{R}^{n_\alpha}}{\text{minimize}} && \mathcal{L}(\mathbf{y}(\boldsymbol{\alpha}_r), \boldsymbol{\alpha}_r) \\ & \text{subject to} && \mathbf{c}(\mathbf{y}(\boldsymbol{\alpha}_r), \boldsymbol{\alpha}_r) \leq 0 \end{aligned}$$

- $\mathbf{y}$  is implicitly defined as a function of  $\boldsymbol{\alpha}_r$  through the ROM equation

$$\boldsymbol{\Phi}^T \mathbf{f}^{int}(\boldsymbol{\Phi} \mathbf{y}) = \boldsymbol{\Phi}^T \mathbf{f}^{ext}$$



## Structural Simulation

- St. Venant-Kirchhoff
- 66,191 tetrahedral elements
- 13,110 nodes, 38,664 dof
- Static simulation with load applied in 10 increments
- Loads: Bending, Twisting, Self-Weight
- ROM size:  $k_{\mathbf{u}} = 5$



## Simulation Results

	Offline (s)	Online (s)	Speedup	Error (%)
HDM	-	750	-	-
ROM	0.38	170	3.96	0.003
ROM-precomp	5,171	0.37	<b>2,051</b>	0.003



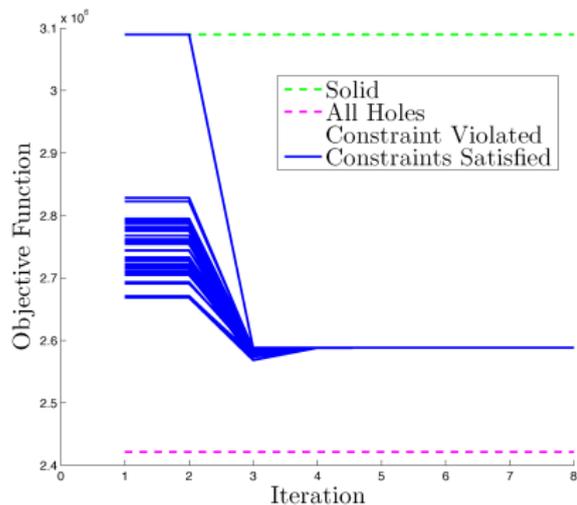
## Optimization Setup

- Minimize structural weight
- Constraint on maximum vertical displacement
- 46 Material Snapshots
  - 45 possible voids
  - volume surrounding all possible voids

Material Snapshots



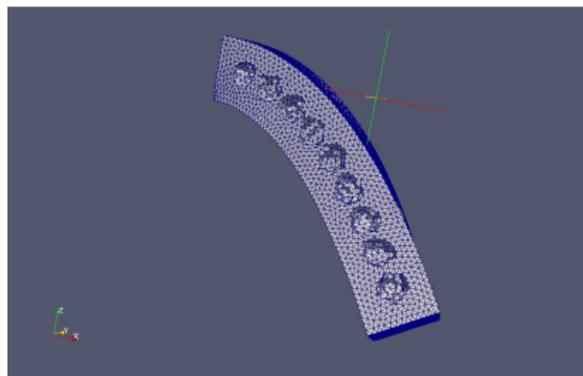
## Optimization Results



Optimization Iterates  
(Location of Voids)



## Optimization Results



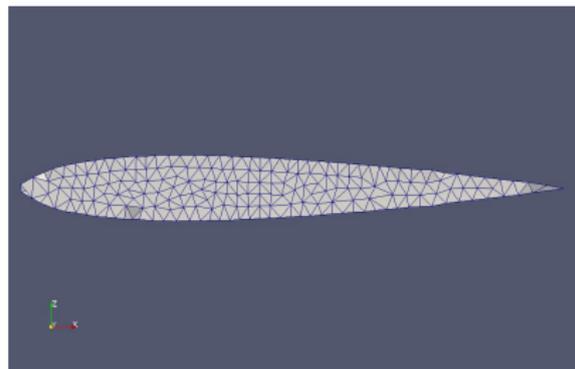
Deformed Configuration (Optimal Solution)

	Initial Guess	Optimal Solution
Structural Weight	$2.776 \times 10^6$	$2.588 \times 10^6$
Constraint Violation	$9.96 \times 10^{-2}$	$1.34 \times 10^{-10}$



## Problem Setup

- St. Venant-Kirchhoff
- 90,799 tetrahedral elements
- 29,252 nodes, 86,493 dof
- Static simulation with load applied in 10 increments
- Loads: Bending (X- and Y- axis), Twisting, Self-Weight
- ROM size:  $k_{\mathbf{u}} = 5$

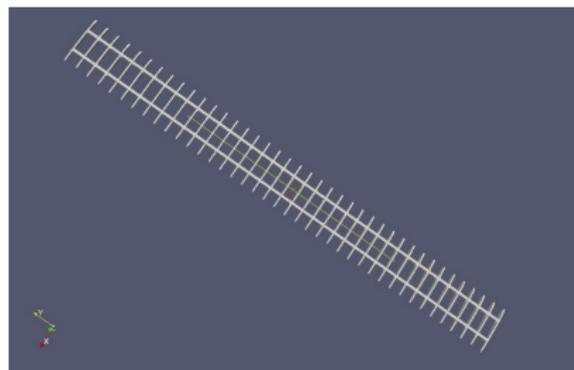


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- St. Venant-Kirchhoff
- 90,799 tetrahedral elements
- 29,252 nodes, 86,493 dof
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40 Ribs



## Problem Setup

- St. Venant-Kirchhoff
- 90,799 tetrahedral elements
- 29,252 nodes, 86,493 dof
- Static simulation with load applied in 10 increments
- Loads: Bending (X- and Y- axis), Twisting, Self-Weight
- ROM size:  $k_{\mathbf{u}} = 5$



## Simulation Results

	Offline (s)	Online (s)	Speedup	Error (%)
HDM	-	811	-	-
ROM	1.01	376	2.16	0.002
ROM-precomp	9,603	1.51	<b>538</b>	1.73



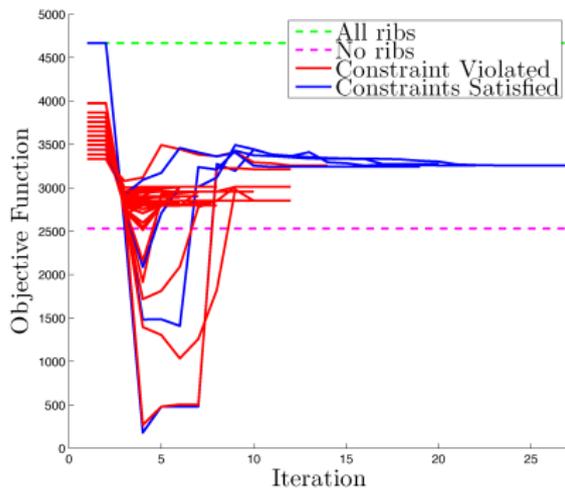
## Optimization Setup

- Minimize structural weight
- Constraint on maximum vertical horizontal displacements
- 41 Material Snapshots
  - 40 possible ribs
  - two spars jointly

Material Snapshots



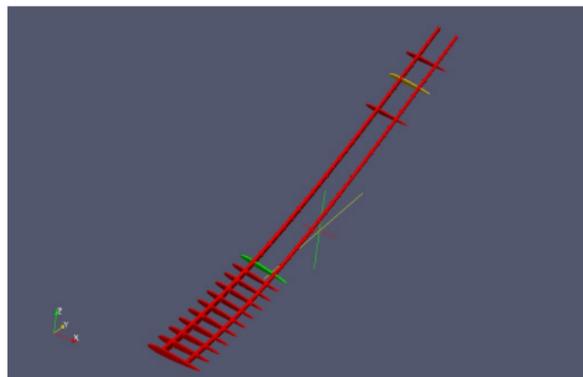
# Optimization Results



Optimization Iterates



## Optimization Results



Deformed Configuration (Optimal Solution)

	Initial Guess	Optimal Solution
Structural Weight	$3.44 \times 10^3$	$3.24 \times 10^3$
Constraint Violation	$4.85 \times 10^{-2}$	$1.19 \times 10^{-16}$



## Conclusion and Future Work

- New method for material topology optimization using reduced-order models
  - $\mathcal{O}(10^3)$  speedup over HDM
- Strongly enforce manufacturability constraints
  - selection of material snapshots and optimization constraints
- Potential to address large problems
- Investigate extending method to more sophisticated topology optimization techniques

