

# Thermomechanical Topology Optimization of Shape-Memory Alloy Structures Using a Transient Bilevel Adjoint Method



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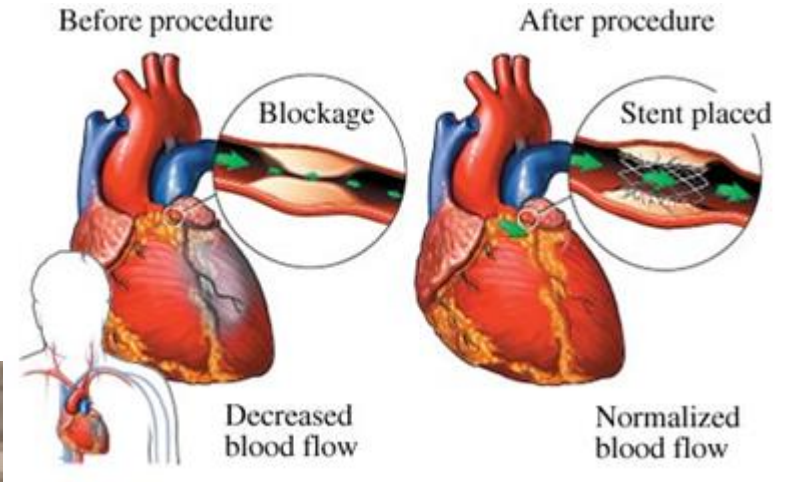
Topology Optimization Webinar  
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# Shape-Memory Alloys: An Overview

- Shape change in response to thermal loads
- Continuous “fluid” motion with infinite degrees of freedom
- Bio-inspired design
- Programmable motion

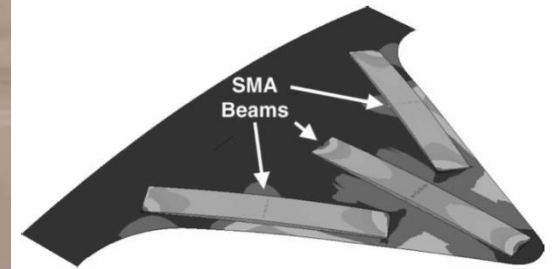


The BioRobotics Institute/Scuola Superiore Sant'Anna 2019



NASA 2018

ANSYS 2015



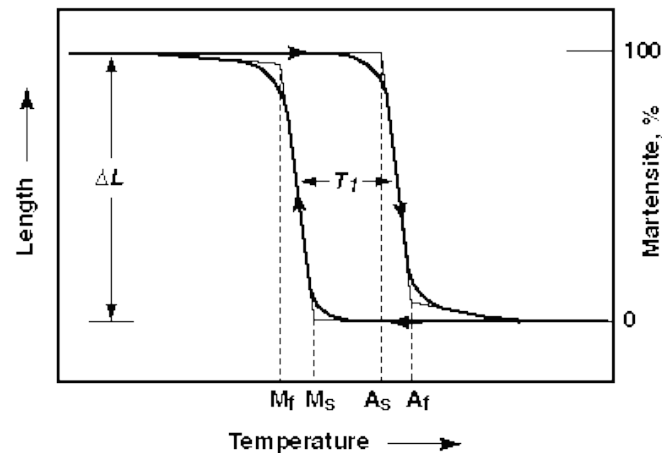
Boeing 2008

- ❖ Key contributions: - Novel topology optimization formulation
- Bi-level adjoint sensitivity formulation

# Phase Transformation

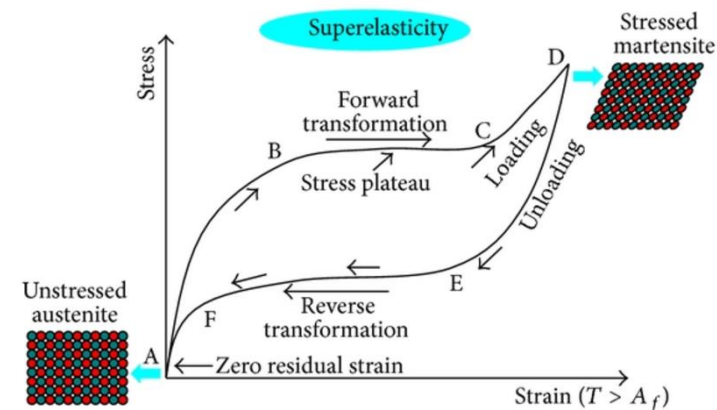
- Inelastic stress and strain response due to phase transformation
- Triggered by latent heat exchange
- Transformation between Martensite and Austenite phases

Two-way shape-memory effect



Lagoudas 2008

Superelasticity



Seo 2015

Lagoudas, DC. *Shape Memory Alloys: Modeling and Engineering Applications*. Berlin, Germany: Springer; 2008.

Seo, Junwon, Young Chan Kim, and Jong Wan Hu. "Pilot study for investigating the cyclic behavior of slit damper systems with recentering shape memory alloy (SMA) bending bars used for seismic restrainers." *Applied Sciences* 5.3 (2015): 187-208.

# Constitutive Model

- Additive decomposition of small strains

$$\boldsymbol{\varepsilon} = \boldsymbol{\varepsilon}^e(\boldsymbol{\sigma}) + \boldsymbol{\varepsilon}^t(\boldsymbol{\sigma}, T, \xi) + \boldsymbol{\varepsilon}^{th}(T)$$

$$\boldsymbol{S} = \boldsymbol{S}^A + \xi(\boldsymbol{S}^M - \boldsymbol{S}^A) = \boldsymbol{S}^A + \xi\Delta\boldsymbol{S}$$

$$\alpha = \alpha^A = \alpha^M$$

$$c = c^A = c^M$$

$$\dot{\boldsymbol{\varepsilon}}^t = \boldsymbol{\Lambda}\dot{\xi}$$

$$\boldsymbol{\Lambda} = \begin{cases} \frac{3}{2}H \frac{\boldsymbol{\sigma}_s}{\sigma_s^{eff}} & \dot{\xi} > 0 \\ H \frac{\boldsymbol{\varepsilon}_{t-r}}{\varepsilon_{t-r}^{eff}} & \dot{\xi} < 0 \end{cases} \quad \text{Transformation tensor}$$

$$d\boldsymbol{\sigma} = \boldsymbol{\mathfrak{L}} : d\boldsymbol{\varepsilon}$$

$$\boldsymbol{\mathfrak{L}} = \begin{cases} \boldsymbol{S}^{-1} - \frac{\boldsymbol{S}^{-1}:\partial_{\boldsymbol{\sigma}}\Phi \otimes \boldsymbol{S}^{-1}:\partial_{\boldsymbol{\sigma}}\Phi}{\partial_{\boldsymbol{\sigma}}\Phi:\boldsymbol{S}^{-1}:\partial_{\boldsymbol{\sigma}}\Phi - \partial_{\xi}\Phi} & \dot{\xi} > 0 \\ \boldsymbol{S}^{-1} - \frac{\boldsymbol{S}^{-1}:\partial_{\boldsymbol{\sigma}}\Phi \otimes \boldsymbol{S}^{-1}:\partial_{\boldsymbol{\sigma}}\Phi}{\partial_{\boldsymbol{\sigma}}\Phi:\boldsymbol{S}^{-1}:\partial_{\boldsymbol{\sigma}}\Phi + \partial_{\xi}\Phi} & \dot{\xi} < 0 \end{cases}$$

# Finite Element Implementation

Global level :

$$\mathbf{R}_{n+1} = \bigwedge_{\text{el}} \left( \sum_{\text{G}^d} (w \mathbf{B}_{\text{G}^d}^T \boldsymbol{\sigma}_{\text{G}^d, n+1}) - \sum_{\text{G}^f} (\varpi \mathbf{N}_{\text{G}^f} \mathbf{p}_{\text{G}^f, n+1}) \right) = 0$$

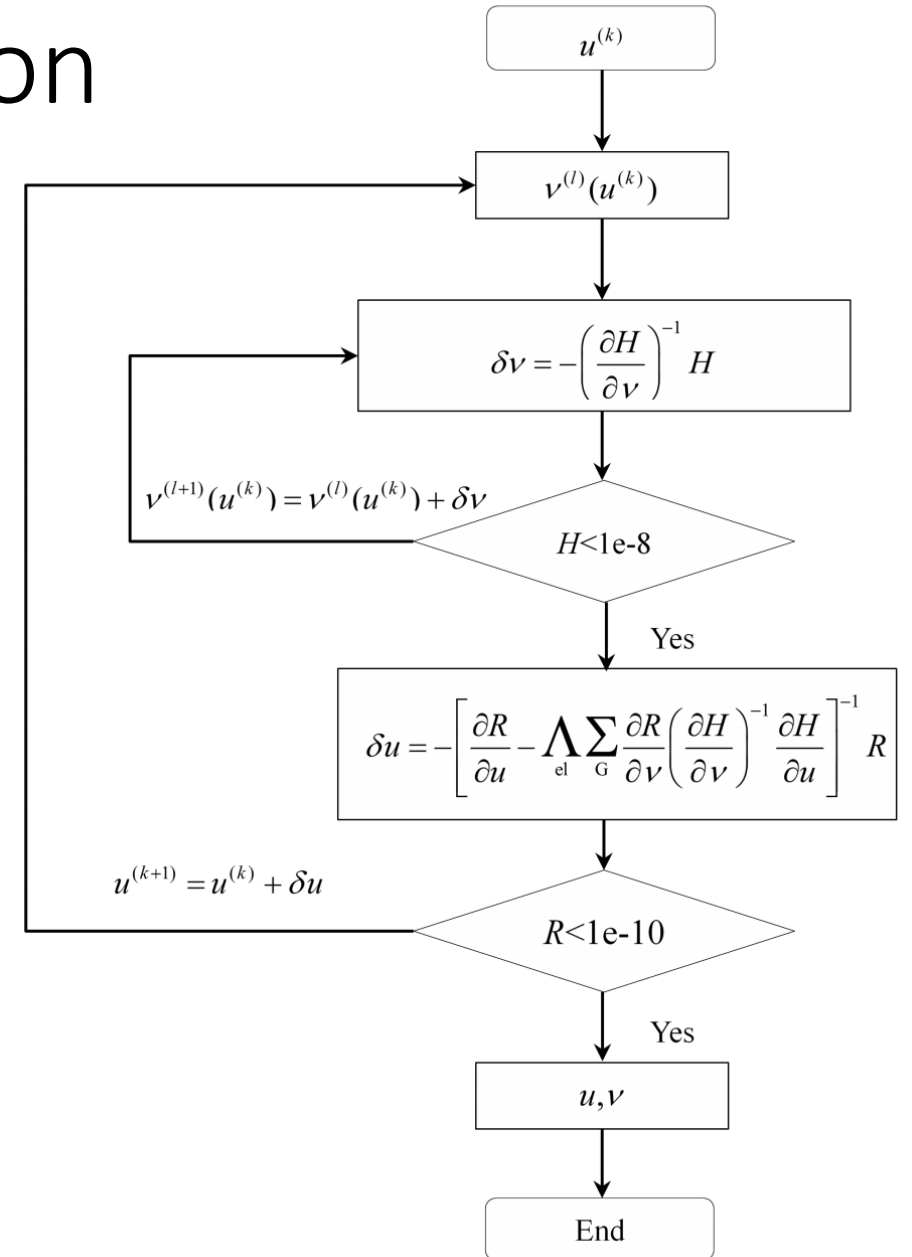
Local level :

$\dot{\xi} \neq 0$ :

$$\mathbf{H}_{n+1} = \begin{cases} H_\xi = \Phi(\boldsymbol{\sigma}_{n+1}, T_{n+1}, \xi_{n+1}) \\ \mathbf{H}_{\boldsymbol{\varepsilon}^t} = \boldsymbol{\varepsilon}_n^t + \boldsymbol{\Lambda}(\xi_{n+1} - \xi_n) - \boldsymbol{\varepsilon}_{n+1}^t \\ H_S = S_n + \Delta S(\xi_{n+1} - \xi_n) - S_{n+1} \\ \mathbf{H}_\boldsymbol{\sigma} = \boldsymbol{\sigma}_n + \mathbf{S}_{n+1}^{-1} : [\mathbf{B} \mathbf{d}_{n+1} - \boldsymbol{\alpha}(T_{n+1} - T_0) - \boldsymbol{\varepsilon}_{n+1}^t] \\ \quad - \mathbf{S}_n^{-1} : [\mathbf{B} \mathbf{d}_n - \boldsymbol{\alpha}(T_n - T_0) - \boldsymbol{\varepsilon}_n^t] - \boldsymbol{\sigma}_{n+1} \end{cases}$$

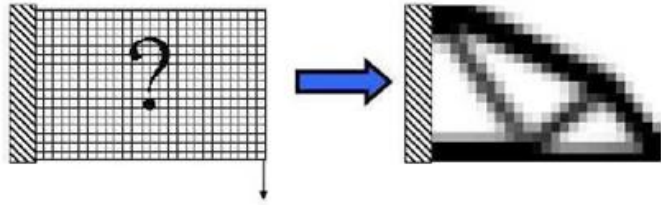
$\dot{\xi} = 0$ :

$$\mathbf{H}_{n+1} = \boldsymbol{\sigma}_n + \mathbf{S}^{-1} : [(\mathbf{B} \mathbf{d}_{n+1} - \boldsymbol{\alpha}(T_{n+1} - T_0) - \boldsymbol{\varepsilon}^t) - (\mathbf{B} \mathbf{d}_n - \boldsymbol{\alpha}(T_n - T_0) - \boldsymbol{\varepsilon}^t)] - \boldsymbol{\sigma}_{n+1} = 0$$



# Optimization Problem Formulation

Two-phase topology optimization:



Two-way shape-memory effect  
(geometric advantage):

$$f_{obj} = \mathbf{L}^T \mathbf{d}$$

Superelasticity (mechanical advantage):

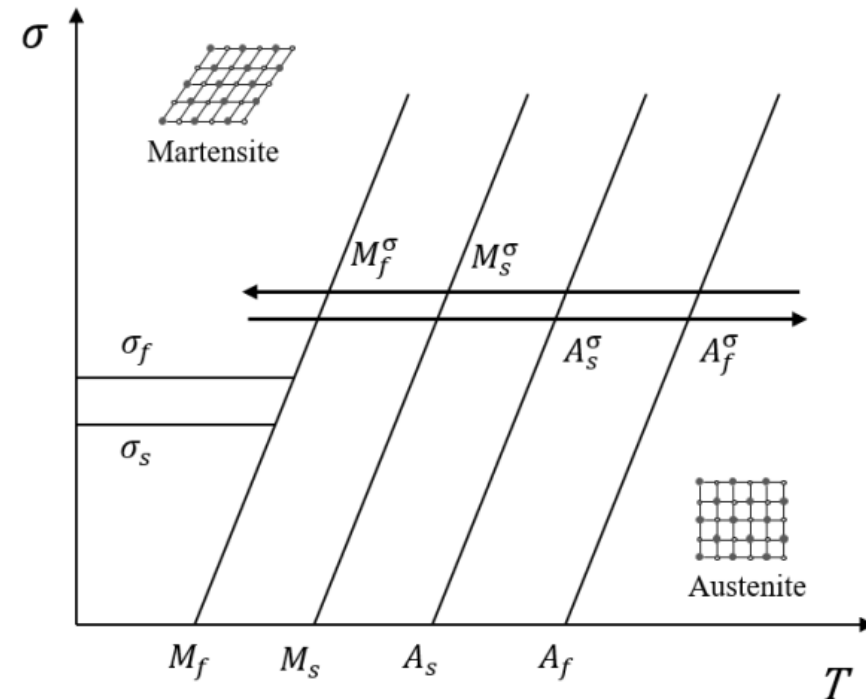
$$f_{obj} = \frac{\mathbf{F}_{out}}{|\mathbf{F}_{in}|} = \frac{\mathbf{L}_{out}^T \mathbf{F}}{|\mathbf{F}_{in}|}$$

Material interpolation scheme:

$$E^{eff} = r^p E^*$$

$$dT^{eff} = r^p dT^*$$

$$V^{eff} = rV^*$$

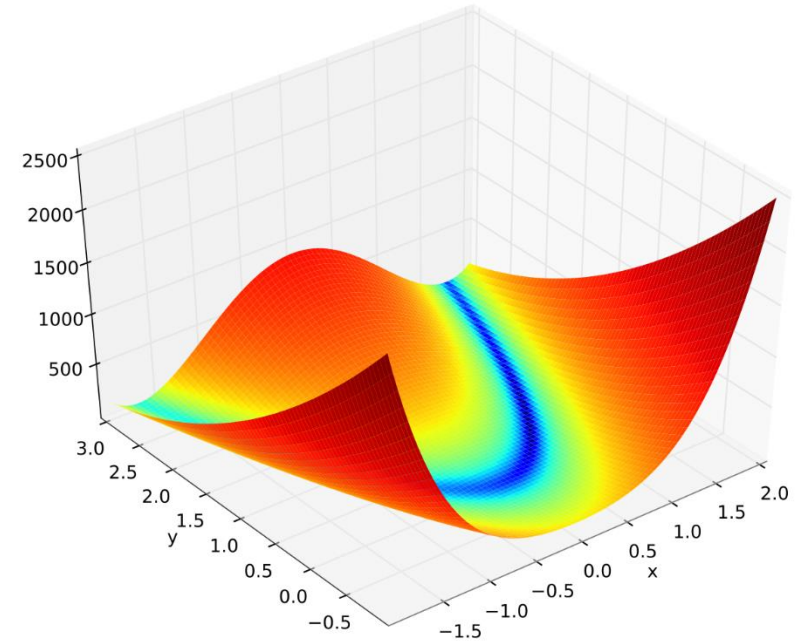




# Adjoint Sensitivity Analysis

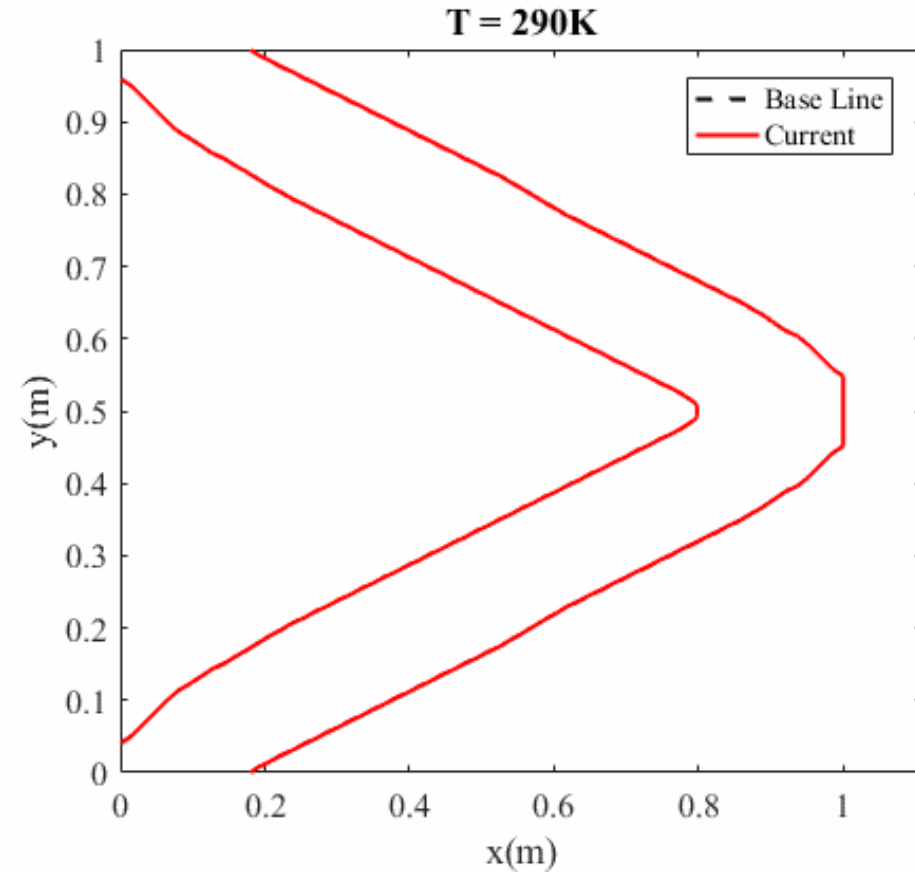
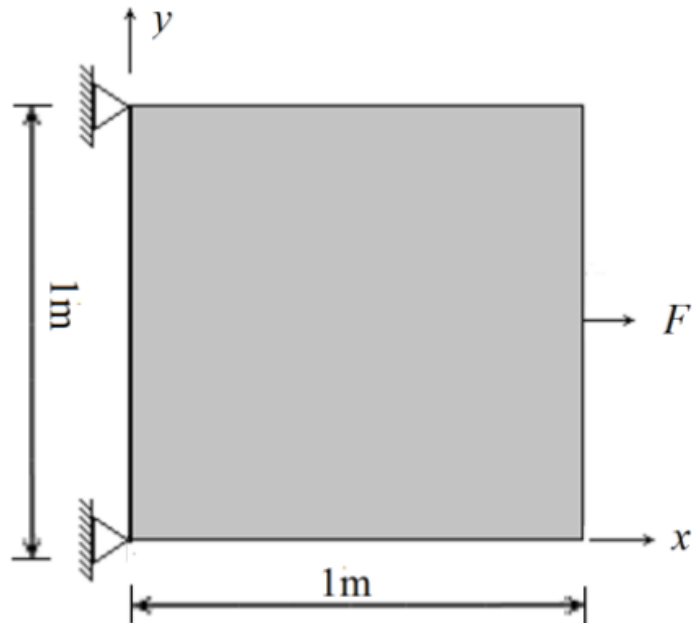
$$n = N_t : \left\{ \begin{array}{l} \lambda_{N_t} = \left[ \frac{\partial \mathbf{R}_{N_t}}{\partial \mathbf{u}_{N_t}} - \bigwedge_{\text{el}} \sum_{\text{G}} \frac{\partial \mathbf{R}_{N_t}}{\partial \boldsymbol{\nu}_{\text{G},N_t}} \left( \frac{\partial \mathbf{H}_{\text{G},N_t}}{\partial \boldsymbol{\nu}_{\text{G},N_t}} \right)^{-1} \frac{\partial \mathbf{H}_{\text{G},N_t}}{\partial \mathbf{u}_{N_t}} \right]^{-\text{T}} \\ \cdot \left[ \bigwedge_{\text{el}} \sum_{\text{G}} \frac{\partial f_{\text{int}}}{\partial \boldsymbol{\nu}_{\text{G},N_t}} \left( \frac{\partial \mathbf{H}_{\text{G},N_t}}{\partial \boldsymbol{\nu}_{\text{G},N_t}} \right)^{-1} \frac{\partial \mathbf{H}_{\text{G},N_t}}{\partial \mathbf{u}_{N_t}} - \frac{\partial f_{\text{int}}}{\partial \mathbf{u}_{N_t}} \right]^{\text{T}} \\ \gamma_{\text{G},N_t} = - \left( \frac{\partial \mathbf{H}_{\text{G},N_t}}{\partial \boldsymbol{\nu}_{\text{G},N_t}} \right)^{-\text{T}} \left( \frac{\partial f_{\text{int}}}{\partial \boldsymbol{\nu}_{\text{G},N_t}} + \lambda_{N_t}^{\text{T}} \frac{\partial \mathbf{R}_{N_t}}{\partial \boldsymbol{\nu}_{\text{G},N_t}} \right)^{\text{T}}, \forall \text{G} \end{array} \right.$$

$$n < N_t : \left\{ \begin{array}{l} \lambda_n = \left[ \bigwedge_{\text{el}} \sum_{\text{G}} \frac{\partial \mathbf{R}_n}{\partial \boldsymbol{\nu}_{\text{G},n}} \left( \frac{\partial \mathbf{H}_{\text{G},n}}{\partial \boldsymbol{\nu}_{\text{G},n}} \right)^{-1} \frac{\partial \mathbf{H}_{\text{G},n}}{\partial \mathbf{u}_n} - \frac{\partial \mathbf{R}_n}{\partial \mathbf{u}_n} \right]^{-\text{T}} \\ \cdot \left\{ \lambda_{n+1}^{\text{T}} \left[ \frac{\partial \mathbf{R}_{n+1}}{\partial \mathbf{u}_n} - \bigwedge_{\text{el}} \sum_{\text{G}} \frac{\partial \mathbf{R}_{n+1}}{\partial \boldsymbol{\nu}_{\text{G},n}} \left( \frac{\partial \mathbf{H}_{\text{G},n}}{\partial \boldsymbol{\nu}_{\text{G},n}} \right)^{-1} \frac{\partial \mathbf{H}_{\text{G},n}}{\partial \mathbf{u}_n} \right] \right. \\ \left. + \bigwedge_{\text{el}} \sum_{\text{G}} \gamma_{\text{G},n+1}^{\text{T}} \left[ \frac{\partial \mathbf{H}_{\text{G},n+1}}{\partial \mathbf{u}_n} - \frac{\partial \mathbf{H}_{n+1}}{\partial \boldsymbol{\nu}_{\text{G},n}} \left( \frac{\partial \mathbf{H}_{\text{G},n}}{\partial \boldsymbol{\nu}_{\text{G},n}} \right)^{-1} \frac{\partial \mathbf{H}_{\text{G},n}}{\partial \mathbf{u}_n} \right] \right\}^{\text{T}} \\ \gamma_{\text{G},n} = - \left( \frac{\partial \mathbf{H}_{\text{G},n}}{\partial \boldsymbol{\nu}_{\text{G},n}} \right)^{-\text{T}} \left( \lambda_n^{\text{T}} \frac{\partial \mathbf{R}_n}{\partial \boldsymbol{\nu}_{\text{G},n}} + \lambda_{n+1}^{\text{T}} \frac{\partial \mathbf{R}_{n+1}}{\partial \boldsymbol{\nu}_{\text{G},n}} + \gamma_{\text{G},n+1}^{\text{T}} \frac{\partial \mathbf{H}_{\text{G},n+1}}{\partial \boldsymbol{\nu}_{\text{G},n}} \right)^{\text{T}}, \forall \text{G} \end{array} \right.$$



# Example Problem 1: TWSME

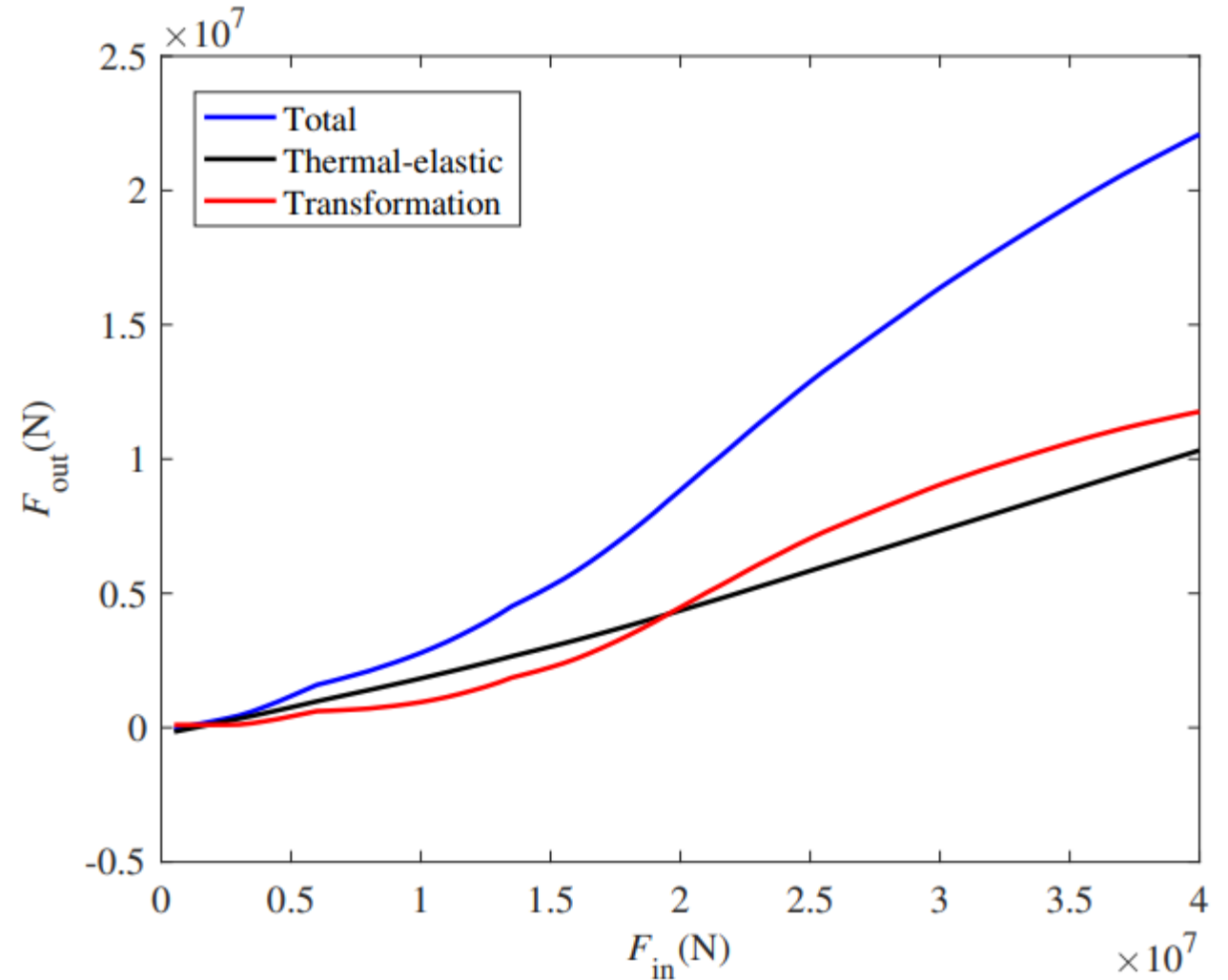
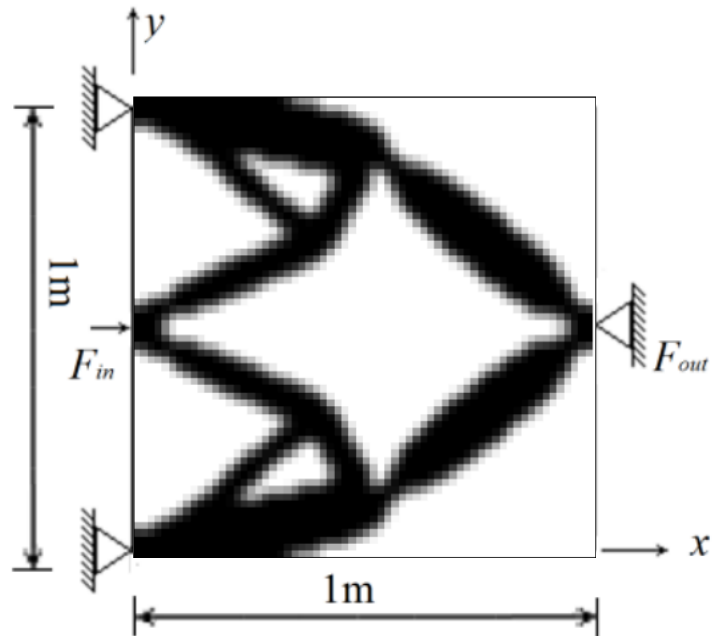
$$\begin{aligned} \min \quad & f_{obj} = -\mathbf{L}^T \mathbf{d} \\ \text{s.t.} \quad & \mathbf{L}^T \tilde{\mathbf{d}} \leq 3 \times 10^{-4} \text{m} \\ & V \leq 0.3 \end{aligned}$$





# Example Problem 2: Superelasticity

$$\begin{aligned} \min \quad & f_{obj} = -\mathbf{F}_{out}/\mathbf{F}_{in}, \\ \text{s.t} \quad & d_{in} < 2 \times 10^{-3} \text{m} \\ & V \leq 0.3 \end{aligned}$$



# Summary & Future Work

- Topology optimization/material interpolation scheme
- Path-dependent, transient adjoint formulation
- 2D examples for TWSME and superelasticity
- Future work:
  - Coupled thermal conductivity (heat diffusion, joule heating)
  - 3D robotic mechanisms; 4D printing

# Acknowledgments



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# Thank you!

Z. Kang and K.A. James, “Thermomechanical topology optimization of shape-memory alloy structures using a transient bi-level adjoint method”. *International Journal for Numerical Methods in Engineering*, 121(11):2558-2580, 2020.

<https://onlinelibrary.wiley.com/doi/abs/10.1002/nme.6319>