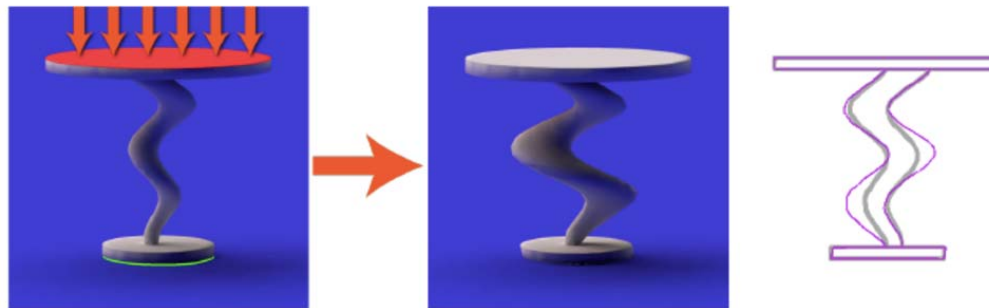


# Direct Shape Optimization for Strengthening 3D Printable Objects



Yahan Zhou, Vangelis Kalogerakis, Rui Wang, Ian R. Grosse  
*University of Massachusetts Amherst*



# Introduction

- Printable 3D objects that need to withstand significant external force.



Coffee Table



Wrench

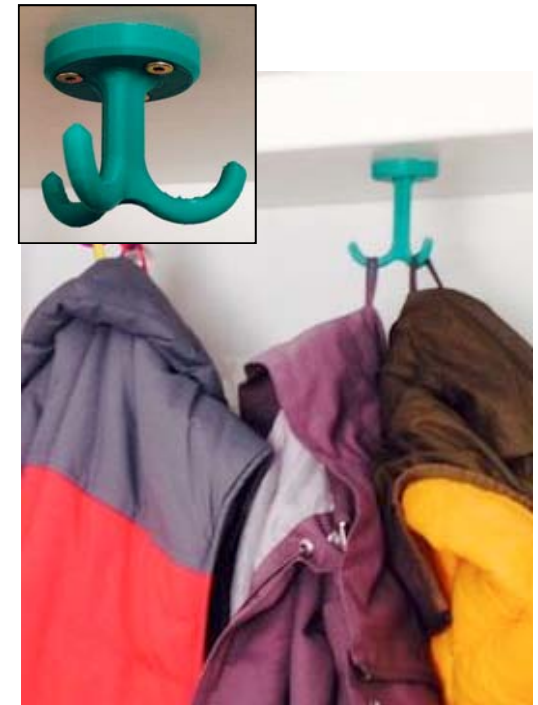


Coat Hanger



# Introduction

- Printable 3D objects that need to withstand significant external force.



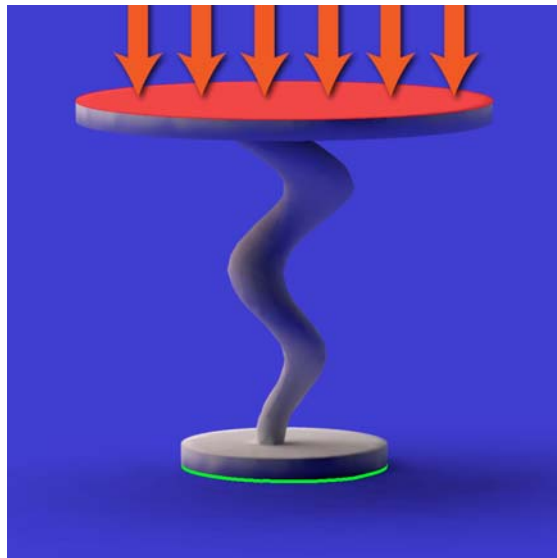
# Introduction

- Given an input 3D shape,



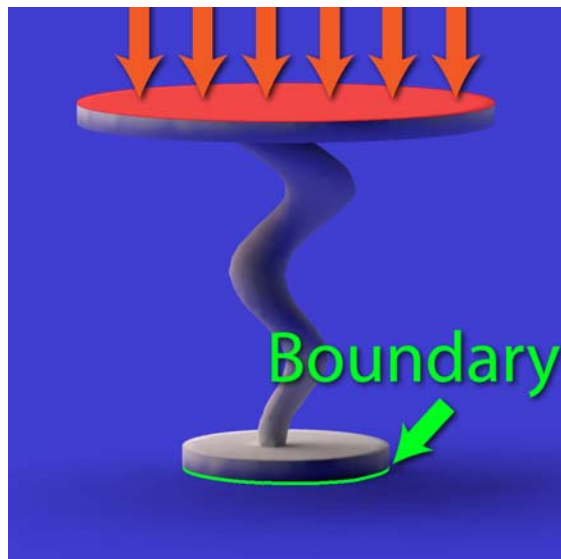
# Introduction

- Given an input 3D shape, the direction and strength of external force,



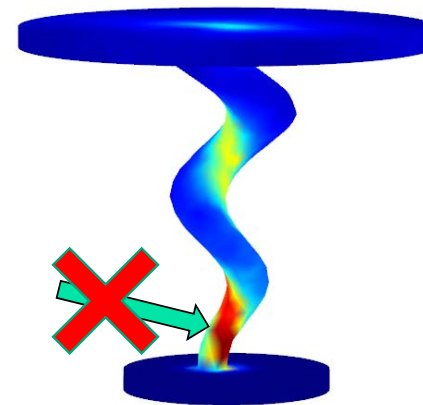
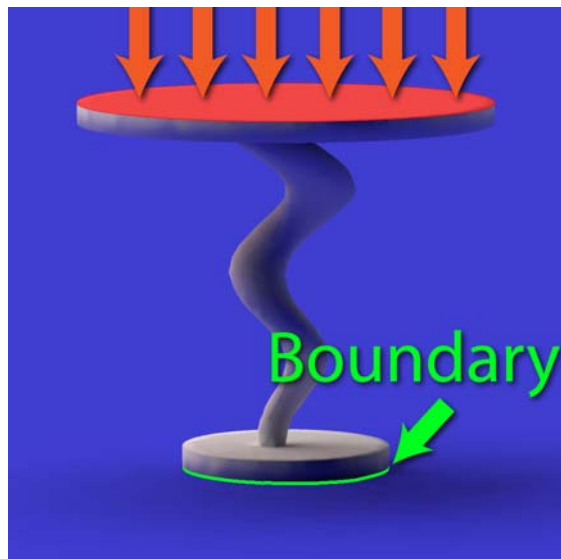
# Introduction

- Given an input 3D shape, the direction and strength of external force, and the boundary,



# Introduction

- The object will deform under such external force. The regions under high stress may break.



Stress Analysis

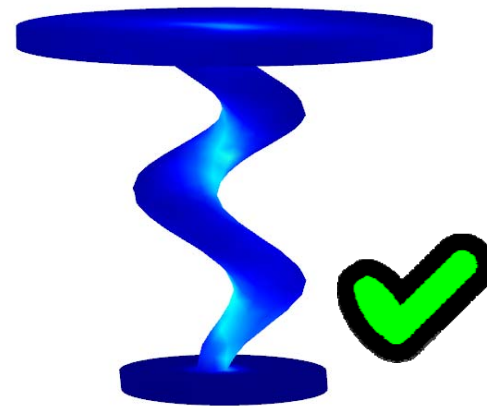


# Introduction

- Our goal is to optimize the shape, such that the resulting shape can successfully withstand the external force, and remain as similar as possible to the input shape.



**Optimized**



Stress Analysis



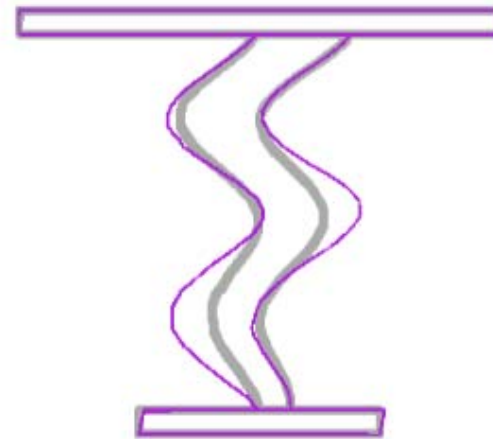


# Introduction

- Silhouette image showing the difference between the original (gray) and optimized (purple) shape.



**Optimized**

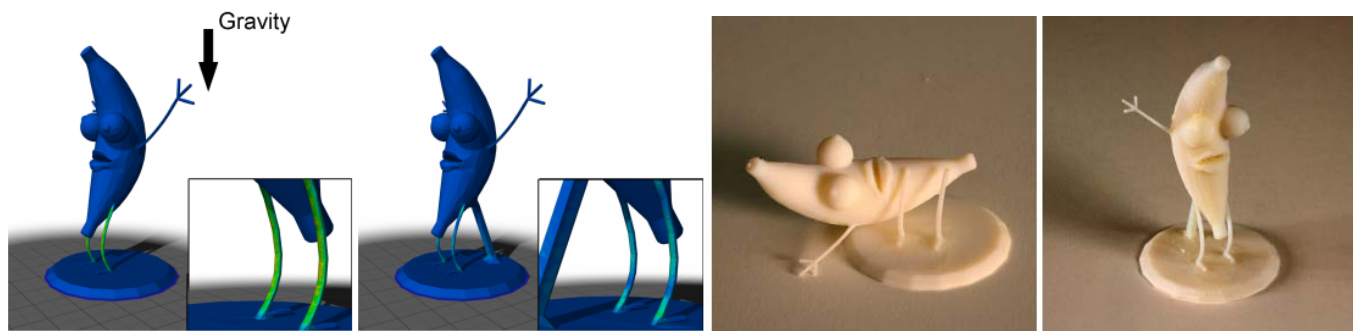


Original vs. Optimized



# Related Work

- Strengthening Printable Objects
  - **Hollowing** [SVB\*12, LSZ\*14, VGB\*14]
  - **Internal skin framing** [WWY\*13]
  - **Support struts** [SVB\*12]
  - **Change printing directions** [HBA13, US13]
  - **Part thickening** [SVB\*12, US13]
  - **Controllable shape design** [MDLW15]

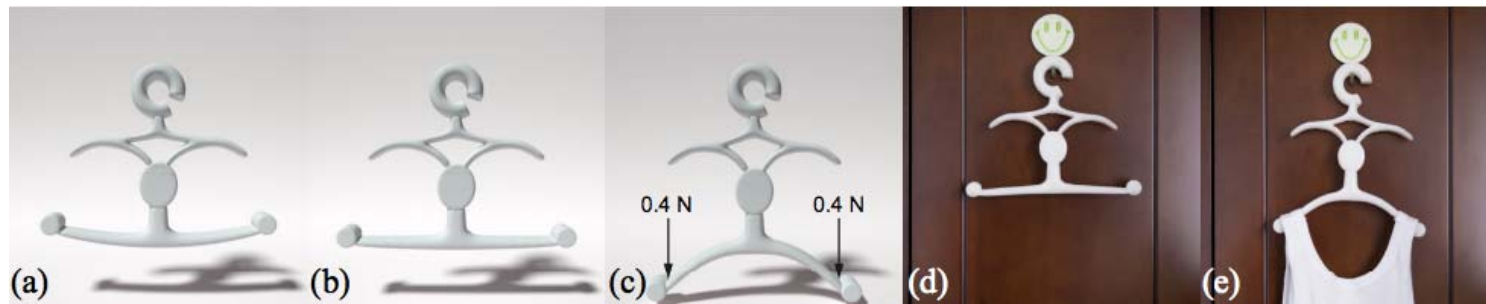


[SVB\*12]



# Related Work

- Shape Optimization in Computer Graphics
  - **Balance** [PWLSH13]
  - **Spinnability** [BWBSH14]
  - **Aggregate mass** [MAB\*15]
  - **Inverse elastic shape design** [CZXZ14]
  - **Microstructures** [PZM\*15, SBR\*15]



[CZXZ14]



# Related Work

- Shape Optimization in Mechanical Engineering
  - **Parametric surface** [HM03, WMC08, BCC\*10]
  - **B-splines/Bezier, subdivision surfaces** [BRC16]
  - **Level-set** [AJT02, AJ08, DMLK13]
  - **Specialized topology modifications** [All12, BS13]
  - **Procedural models** [BR88, RG92, Tor93]

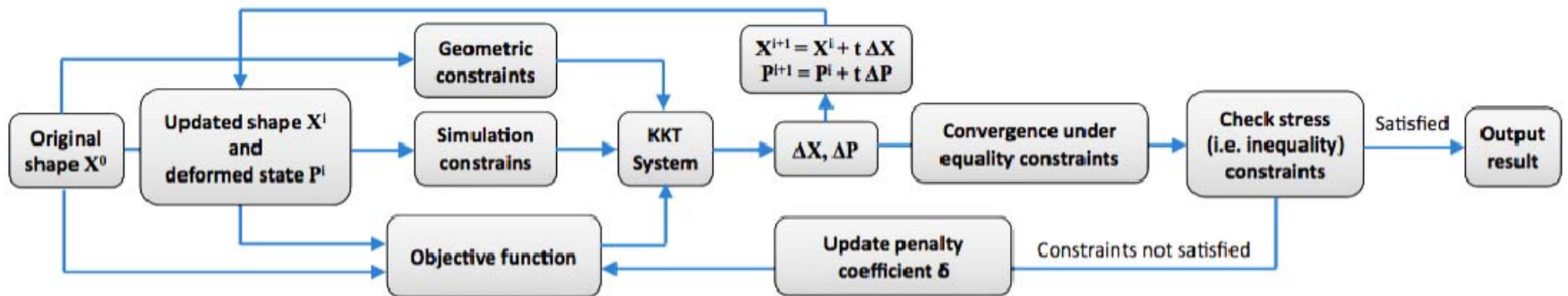


[All12]

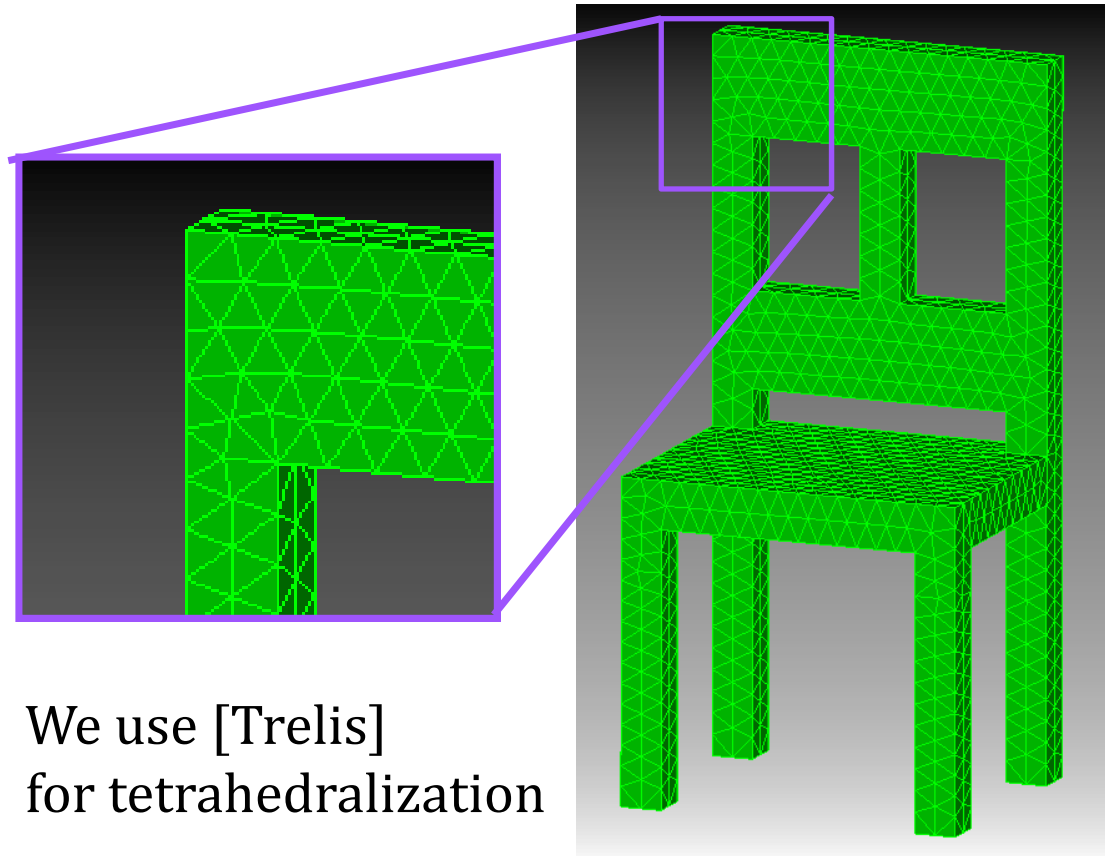


# Contributions

- A general, extensible method to optimize 3D shapes under physical and geometric constraints.
- Operates directly on the input mesh.
- Integrated physics simulation with optimizer.
  - Derivations of analytic gradient and Hessian



# Overview



We use [Trelis]  
for tetrahedralization



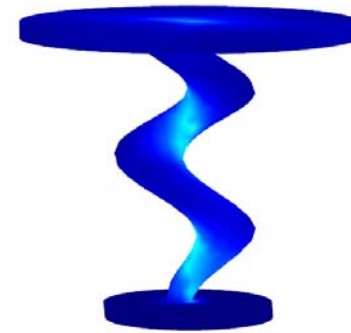
# Overview



$X_0$   
Reference State



$X$   
Rest State



$P$   
Deformed State

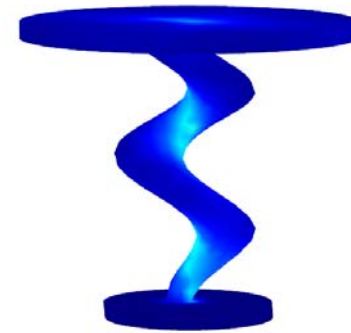
# Overview



$X_0$   
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$P$   
Deformed State



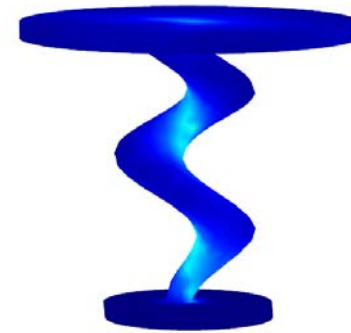
# Overview



$X_0$   
Reference State



$X$   
Rest State



$P$   
Deformed State

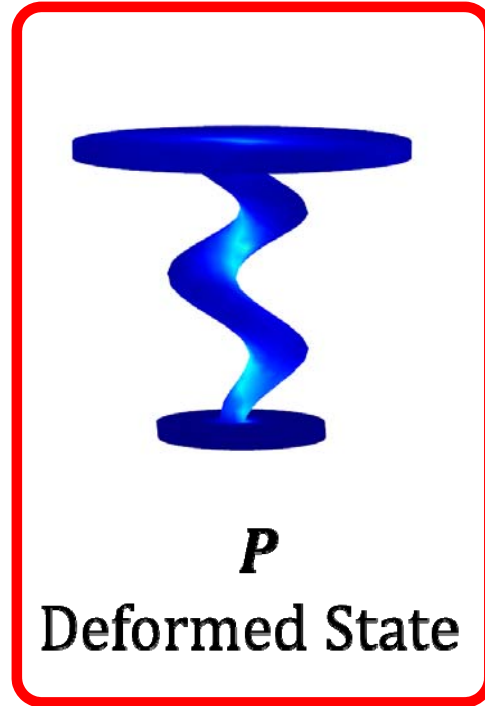
# Overview



$X_0$   
Reference State



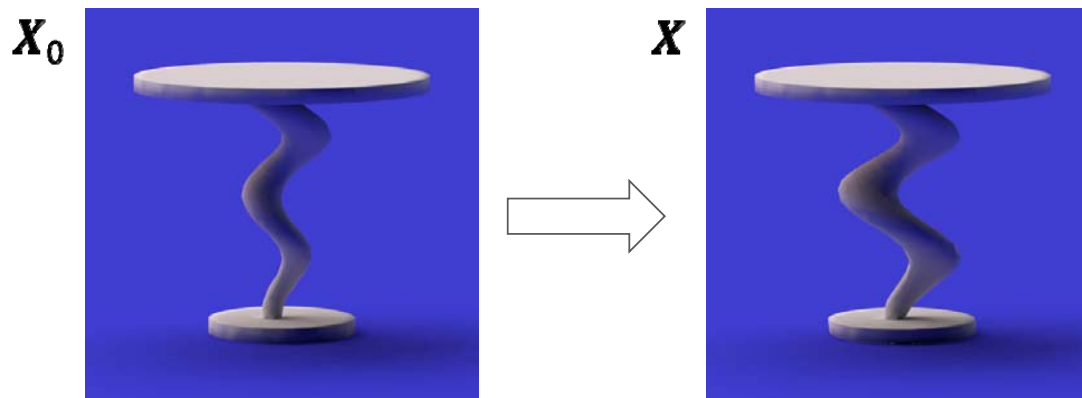
$X$   
Rest State



$P$   
Deformed State

# Formulation – Constrained Optimization

Solve for rest state  $X$  that minimize an objective function while satisfying the given constraints.



Objective:  $\operatorname{argmin}_X D(X, X^0)$

Constraints:  $\forall v \in B: x_v = p_v, \forall v \notin B: f_v(X, P) = 0$  **Simulation**  
 $\forall t: \hat{\sigma}_t^2(X, P) < C$  **Stress**  
 $g(X, X^0) = 0$  **Geometric**

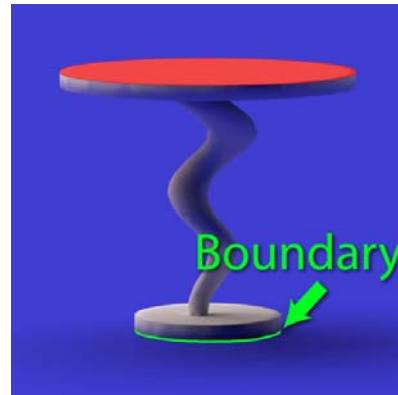


# 1. Simulation Constraints



$X_0$

Reference State



$$\forall v \in B: \underline{x_v} = p_v, \forall v \notin B: f_v(X, P) = 0$$

**Boundary Conditions**

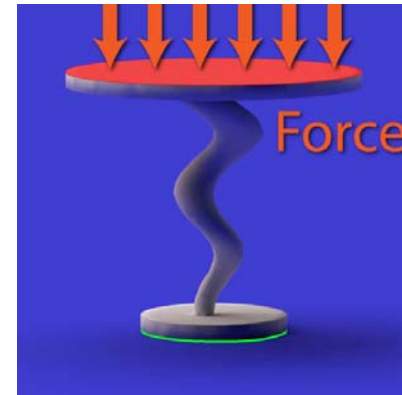
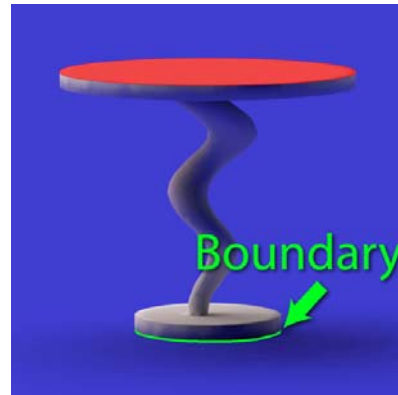


# 1. Simulation Constraints



$X_0$

Reference State



$$\forall v \in B: \mathbf{x}_v = \mathbf{p}_v, \forall v \notin B: \mathbf{f}_v(\mathbf{X}, \mathbf{P}) = \mathbf{0}$$

**Force Equilibrium**

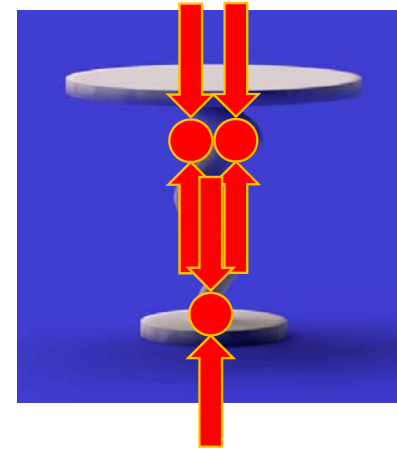
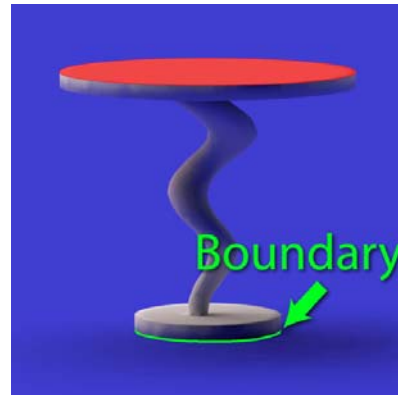


# 1. Simulation Constraints



$X_0$

Reference State



$$\forall v \in B: \mathbf{x}_v = \mathbf{p}_v, \forall v \notin B: \mathbf{f}_v(\mathbf{X}, \mathbf{P}) = \mathbf{0}$$

**Force Equilibrium**



# Elasticity Model



***X***  
Rest State



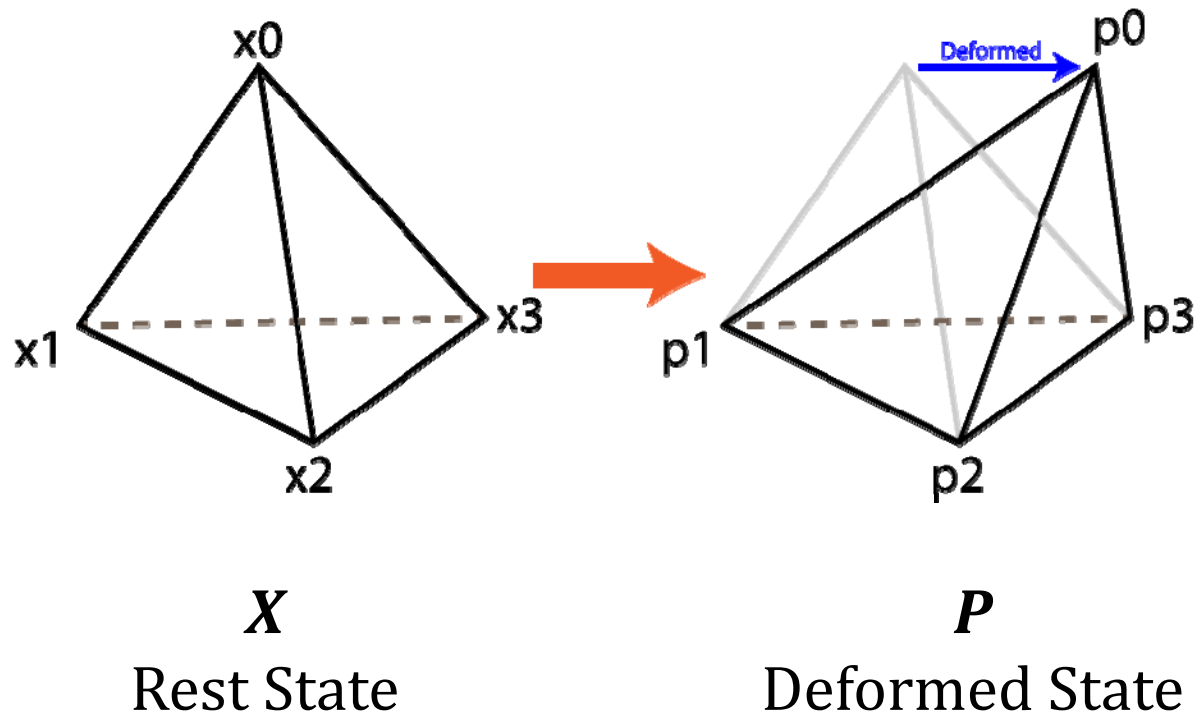
***P***  
Deformed State



Restoration Force

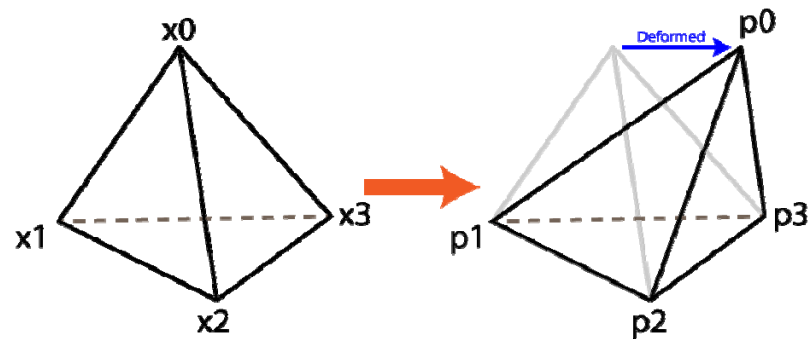


# Elasticity Model





# Elasticity Model



$$\mathbf{f}_{t,v}(\mathbf{X}, \mathbf{P}) = \frac{\partial U_t(\mathbf{X}, \mathbf{P})}{\partial \mathbf{p}_v}$$

Strain Energy: 
$$U_t(\mathbf{X}, \mathbf{P}) = \frac{V_t}{2} \boldsymbol{\varepsilon}_t : \mathbf{E} \boldsymbol{\varepsilon}_t$$

Strain tensor

Material tensor

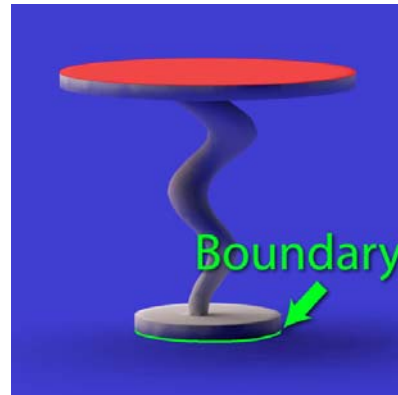


# 1. Simulation Constraints



$X_0$

Reference State



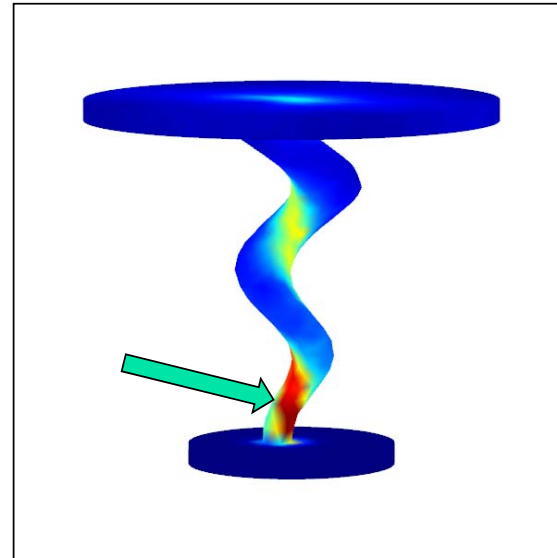
$$\forall v \in B: \underline{x_v = p_v}, \forall v \notin B: \underline{f_v(X, P) = 0}$$

**Boundary Conditions**

**Force Equilibrium**



## 2. Stress Constraints



$$\forall t: \hat{\sigma}_t^2(X, P) < C$$

**von Mises Stress**

**Material's Yield Strength**



# 3. Geometric Constraints

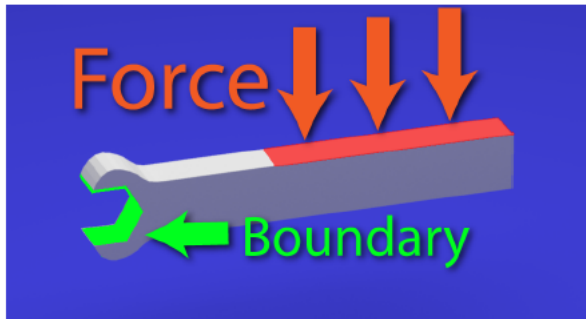
- **Symmetry Constraints:**  $\mathbf{S}_m \mathbf{x}_i = \mathbf{x}_j$
- **Interior Uniformity Constraints:**  $\mathbf{x}_i = \frac{1}{|\mathbb{N}(i)|} \sum_{j \in \mathbb{N}(i)} \mathbf{x}_j$
- **User-Defined Constraints**

- All in the form of linear, equality constraints:

$$\mathbf{g}(\mathbf{X}, \mathbf{X}^0) = 0$$



# Symmetry Constraints



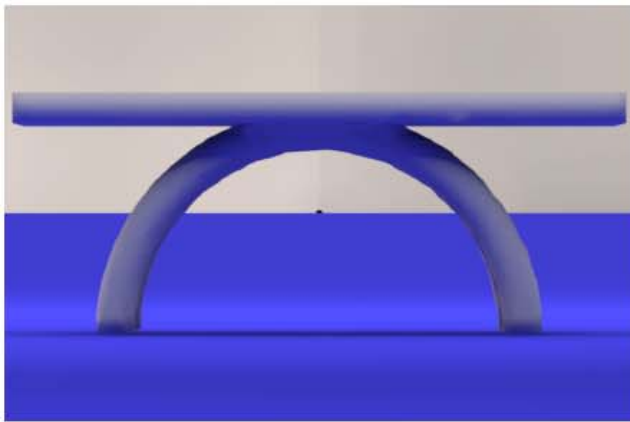
(a) original shape



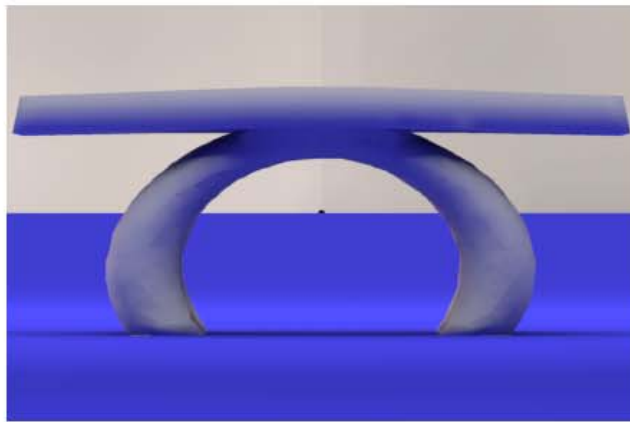
$$\mathbf{S}_m \mathbf{x}_i = \mathbf{x}_j$$



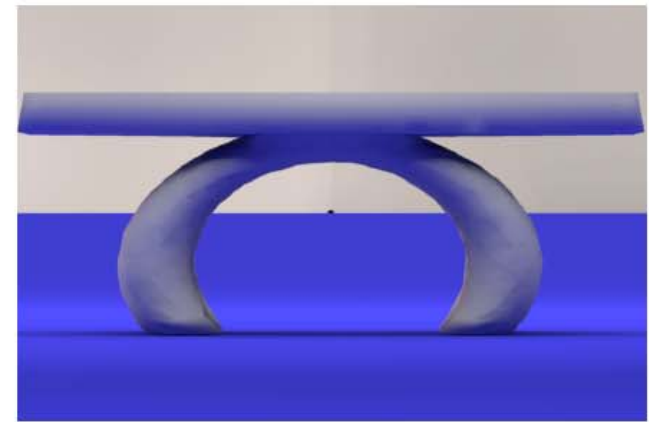
# User-Defined Constraints



(a) original shape



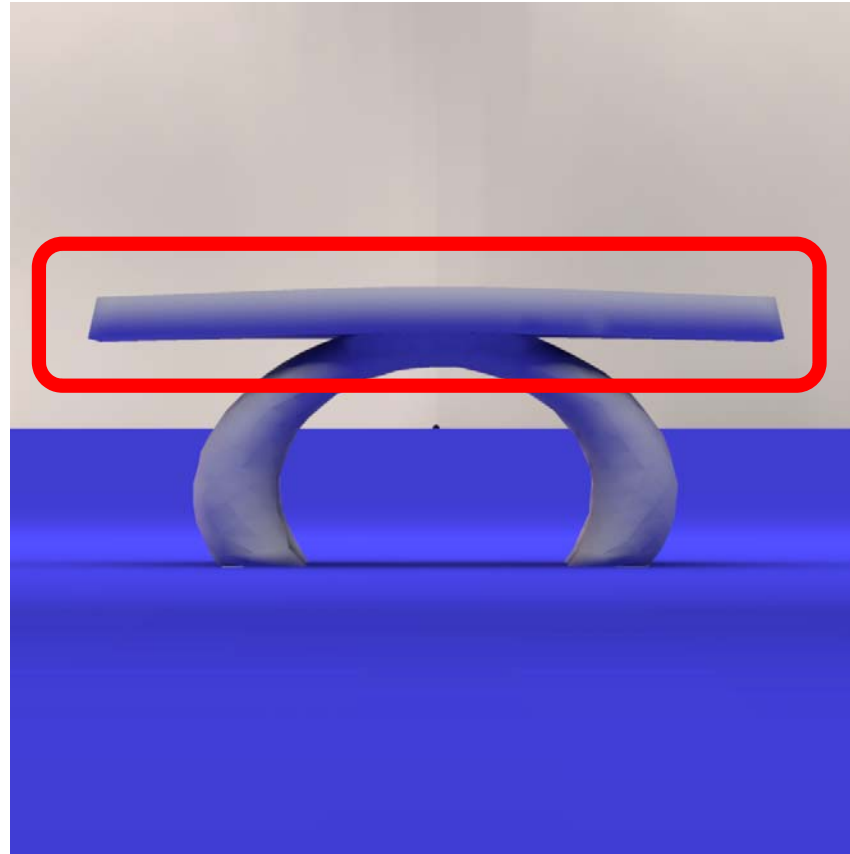
(b) w/o user constraints



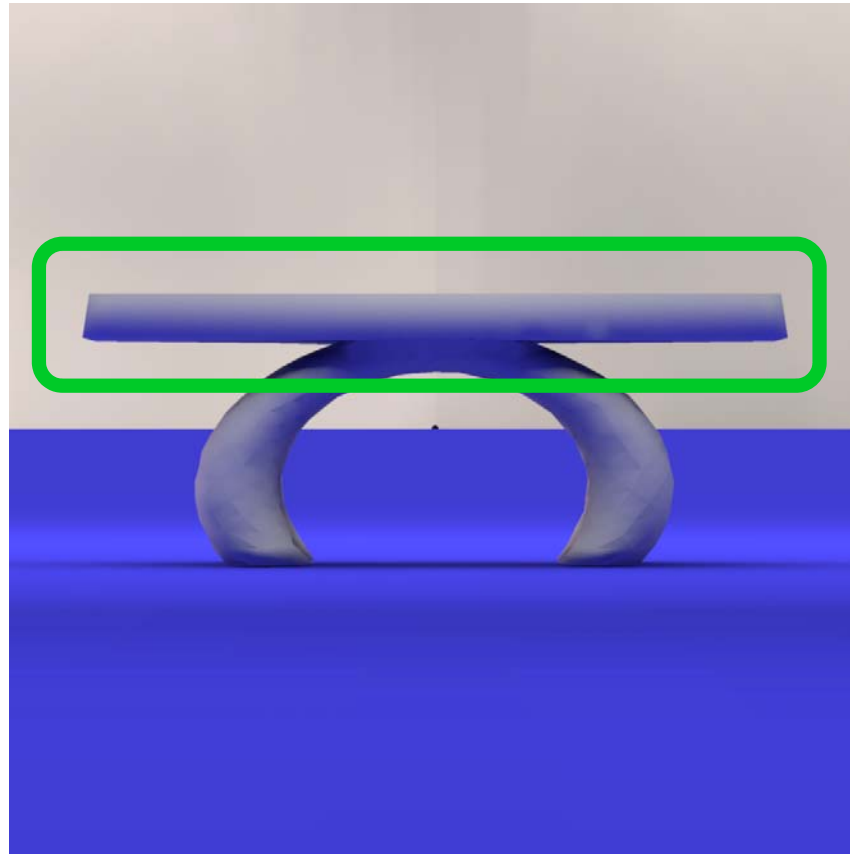
(c) with user constraints



# Without User-Defined Constraints



# With User-Defined Constraints





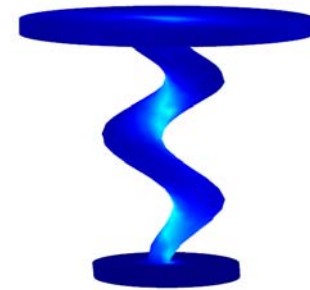
# Objective Function



$X_0$



$X$



$P$

Constraints:  $\forall v \in B: x_v = p_v, \forall v \notin B: f_v(X, P) = 0$  Simulation  
 $\forall t: \hat{\sigma}_t^2(X, P) < C$  Stress  
 $g(X, X^0) = 0$  Geometric

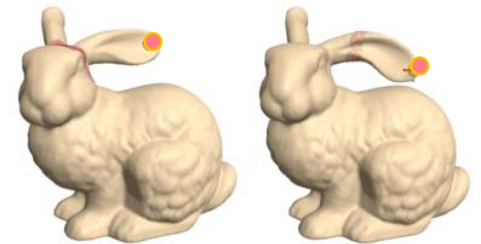


# Objective Function

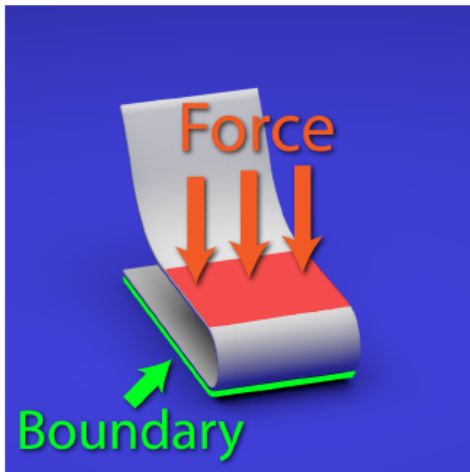
Objective:  $\operatorname{argmin}_{\mathbf{X}} D(\mathbf{X}, \mathbf{X}^0)$

$$D(\mathbf{X}, \mathbf{X}^0) = w_1 D_{intrinsic}(\mathbf{X}, \mathbf{X}^0) + w_2 D_{extrinsic}(\mathbf{X}, \mathbf{X}^0)$$

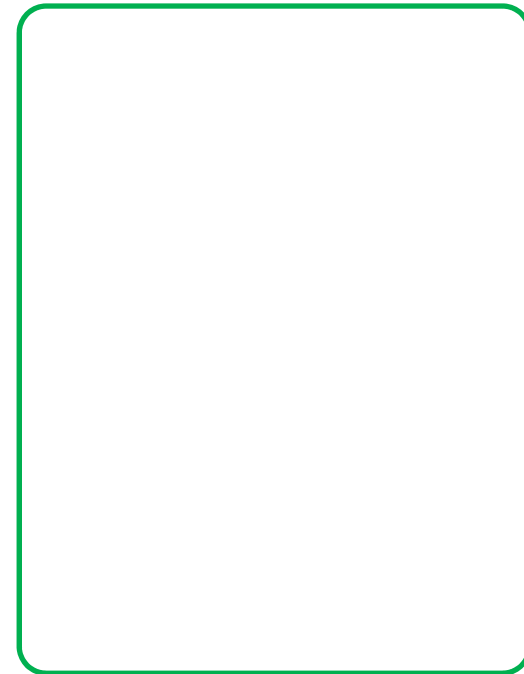
- **Extrinsic:** L2 distance between vertices
  - Preserves overall shape
- **Intrinsic:** transformed surface Laplacians
  - Preserves surface details and smoothness
  - [Sorkine et al. 2004]



# Objective Function



(a) original shape



# Solving the Optimization

Objective:  $\operatorname{argmin}_{\mathbf{X}} D(\mathbf{X}, \mathbf{X}^0)$

Constraints:  $\forall v \in B: \mathbf{x}_v = \mathbf{p}_v, \forall v \notin B: \mathbf{f}_v(\mathbf{X}, \mathbf{P}) = \mathbf{0}$  **Simulation**  
**Stress**  
**Geometric**

**→**  $\forall t: \hat{\sigma}_t^2(\mathbf{X}, \mathbf{P}) < C$   
 $g(\mathbf{X}, \mathbf{X}^0) = 0$

- **Stress Constraints are Inequality constraints:**
  - Use the penalty method.



# The Penalty Method

$$h(\cdot) = \min(0, \cdot)^2$$

Objective:  $\operatorname{argmin}_{\mathbf{X}} D(\mathbf{X}, \mathbf{X}^0) + \delta \cdot \sum_t h(C - \hat{\sigma}_t^2(\mathbf{X}, \mathbf{P}))$

Constraints:  $\forall v \in B: \mathbf{x}_v = \mathbf{p}_v, \forall v \notin B: \mathbf{f}_v(\mathbf{X}, \mathbf{P}) = \mathbf{0}$

$$g(\mathbf{X}, \mathbf{X}^0) = 0$$

- $h$  is a penalty function
- $\delta$  is the penalty weight
  - We start from a small weight and progressively increase it across iterations.



# KKT System and Newton's Method

Objective:  $\operatorname{argmin}_{\mathbf{X}} D(\mathbf{X}, \mathbf{X}^0) + \delta \cdot \sum_t h(C - \hat{\sigma}_t^2(\mathbf{X}, \mathbf{P}))$

Constraints:  $\forall v \in B: \mathbf{x}_v = \mathbf{p}_v, \forall v \notin B: \mathbf{f}_v(\mathbf{X}, \mathbf{P}) = \mathbf{0}$

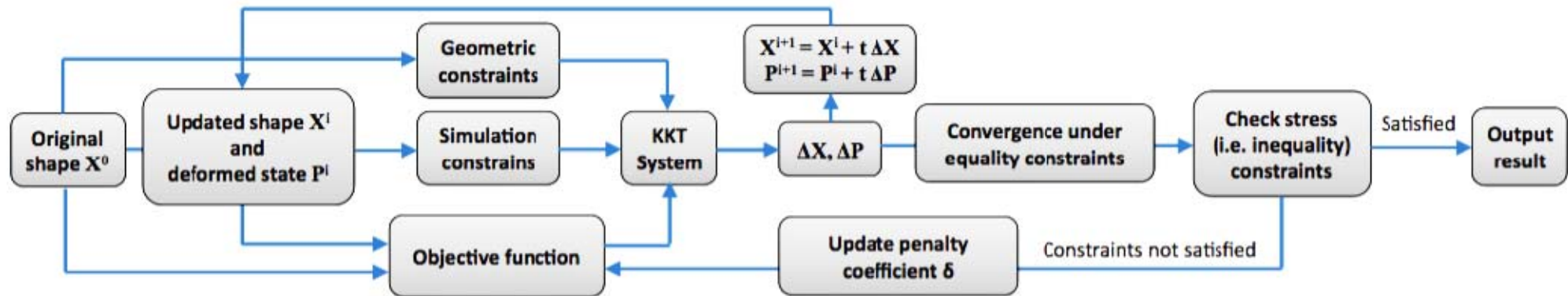
$$\mathbf{g}(\mathbf{X}, \mathbf{X}^0) = \mathbf{0}$$

$$\begin{bmatrix} H & J^T \\ J & 0 \end{bmatrix} \begin{bmatrix} \Delta \mathbf{x} \\ \mathbf{w} \end{bmatrix} = - \begin{bmatrix} \mathbf{g} \\ \mathbf{b} \end{bmatrix}$$

- $\mathbf{g}, H$ : gradient and Hessian of the objective
- $\mathbf{b}, J$ : function value and Jacobian of constraints



# Algorithm Diagram



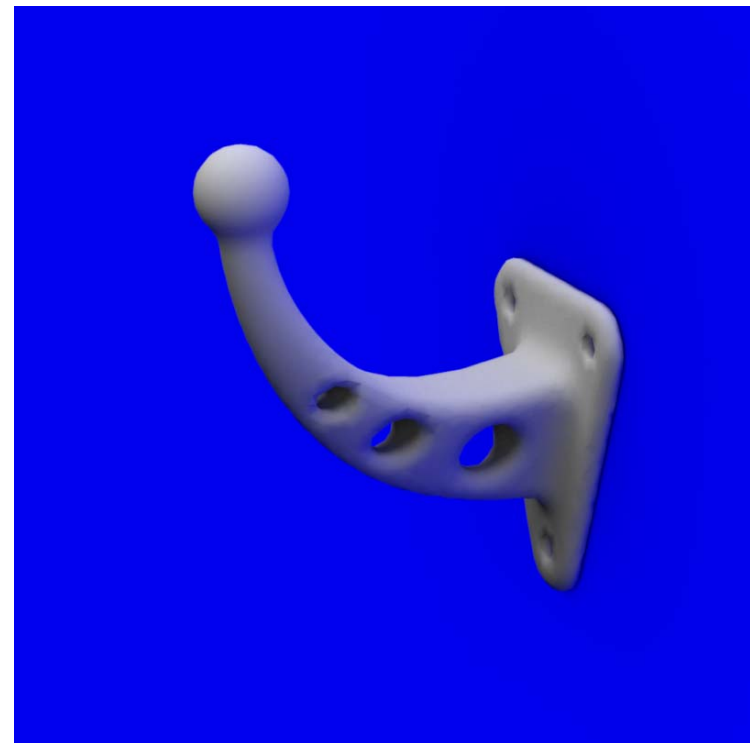
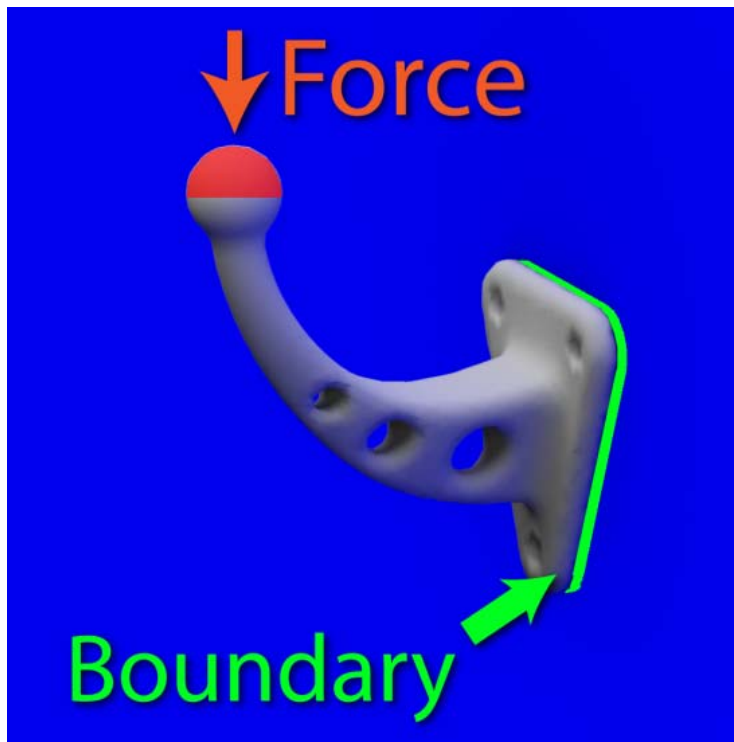
- Source code, data, and supplemental materials available for download from our paper website.



# Results

- Coat Hanger Example:

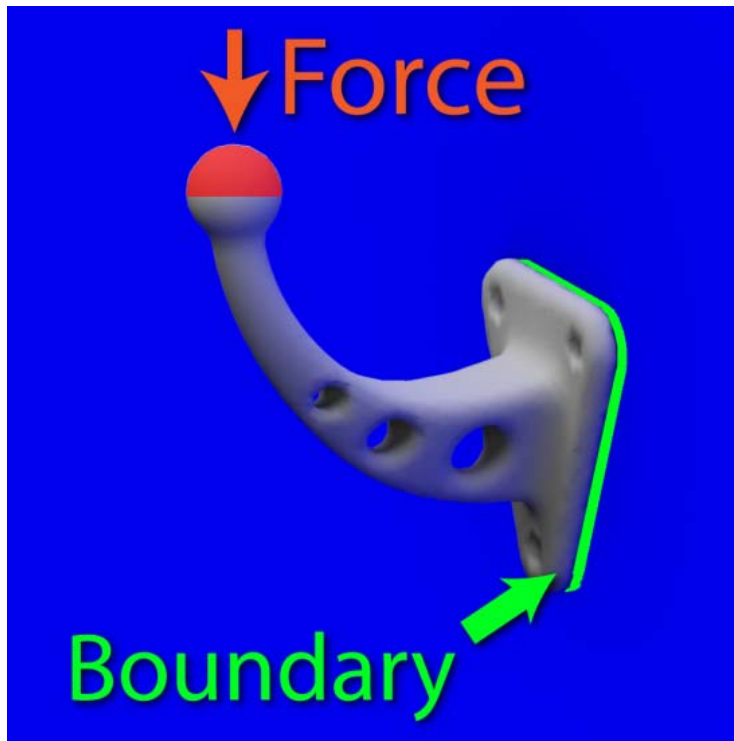
Endure 50% more force





# Results

- Coat Hanger Example:

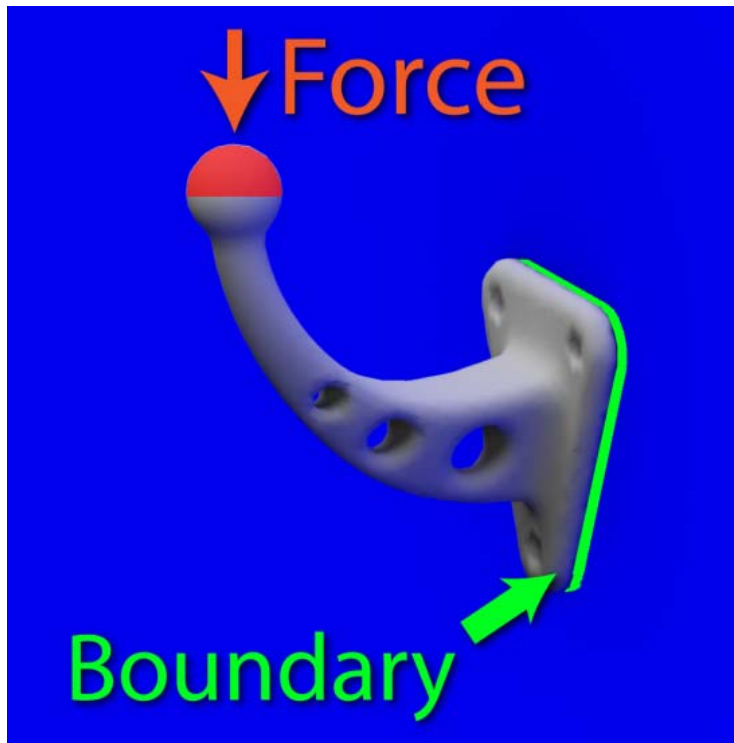


100% more

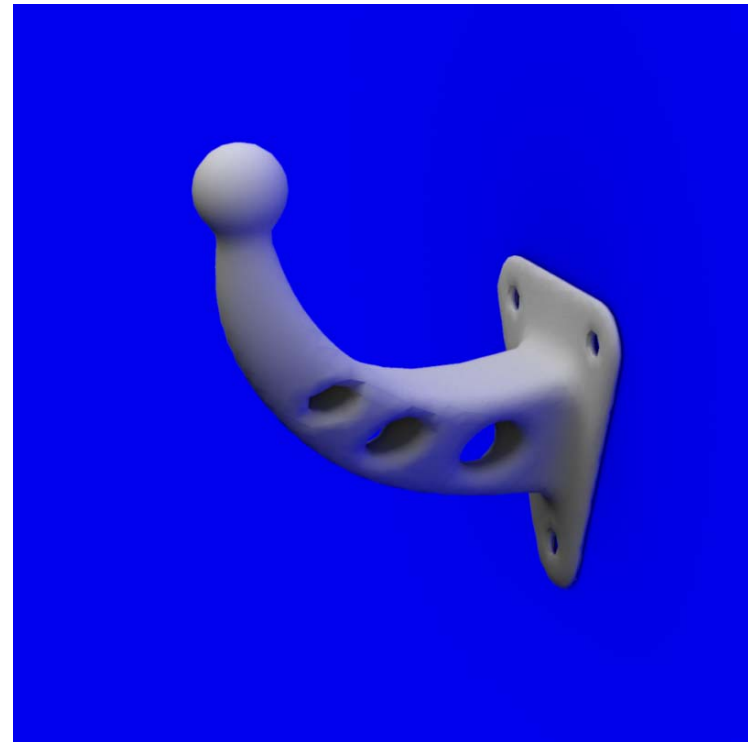


# Results

- Coat Hanger Example:

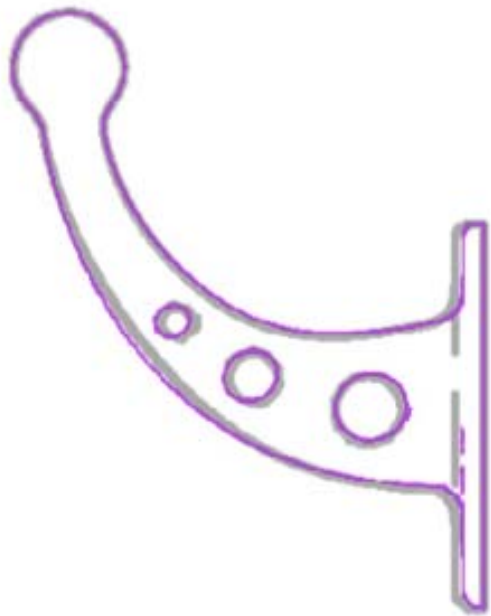


200% more

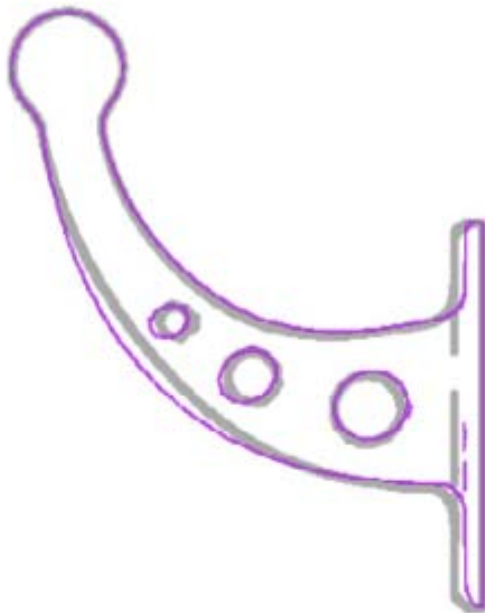


# Results

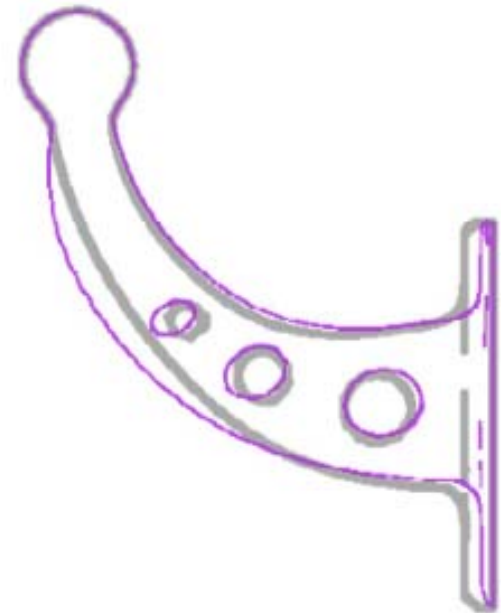
- Coat Hanger Example:



50% more



100% more



200% more



# Comparison with Local Thickening

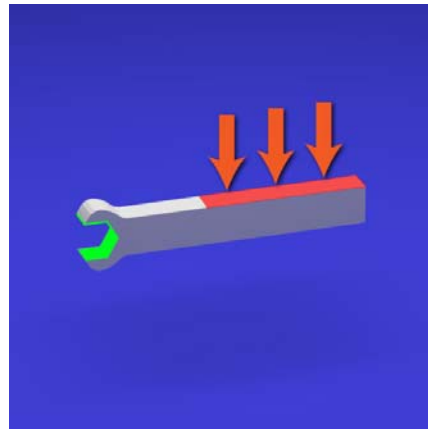
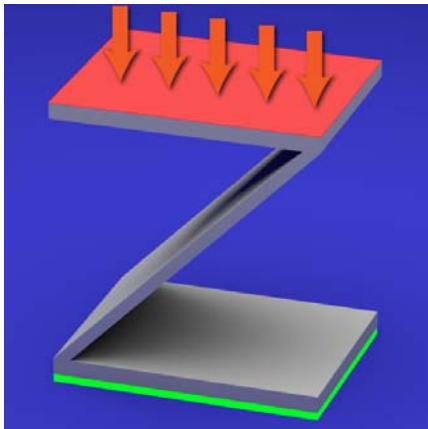
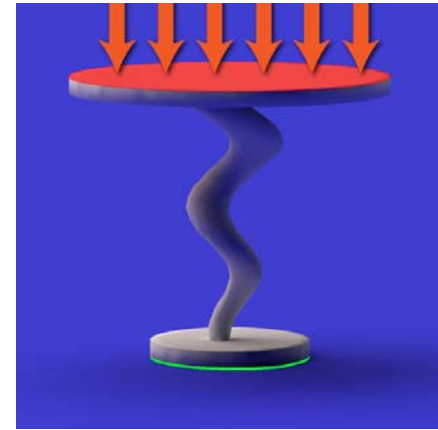
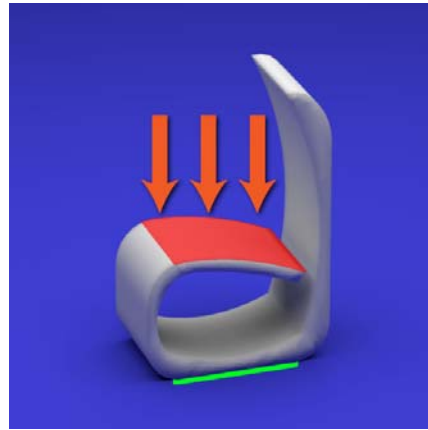
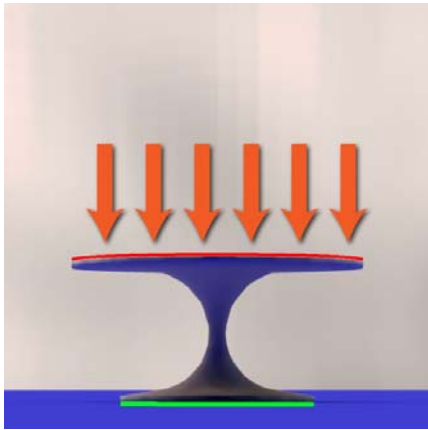


(a) Original shape

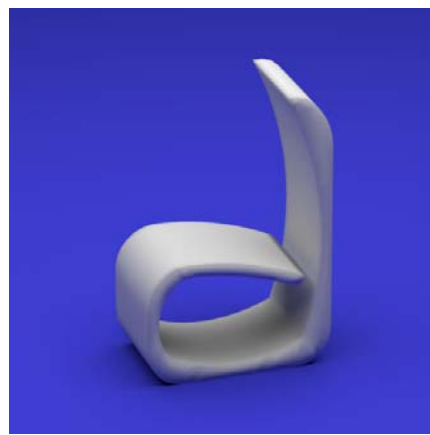


(b) Our method  
**20% less material**  
**Better at preserving**  
**surface features**

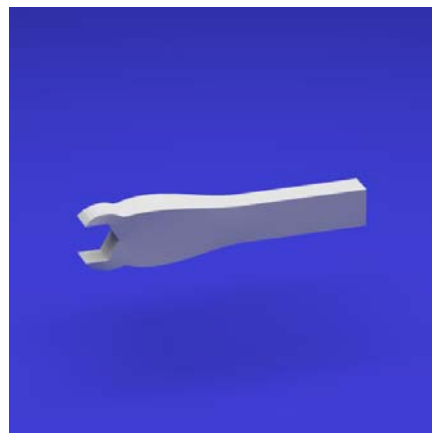
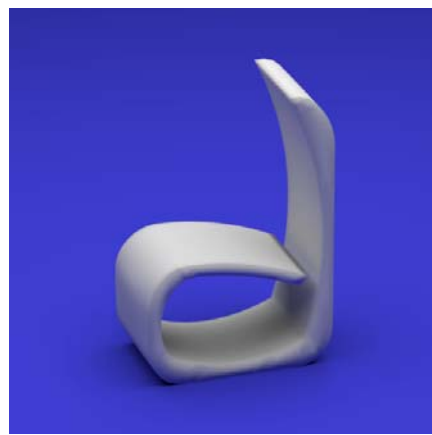
# Gallery of Results – Force and Boundary



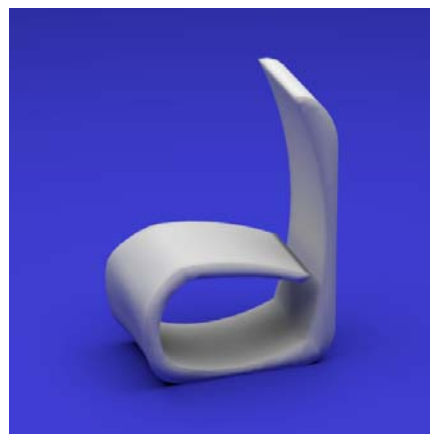
# Gallery of Results – 50% More Force



# Gallery of Results – 100% More Force

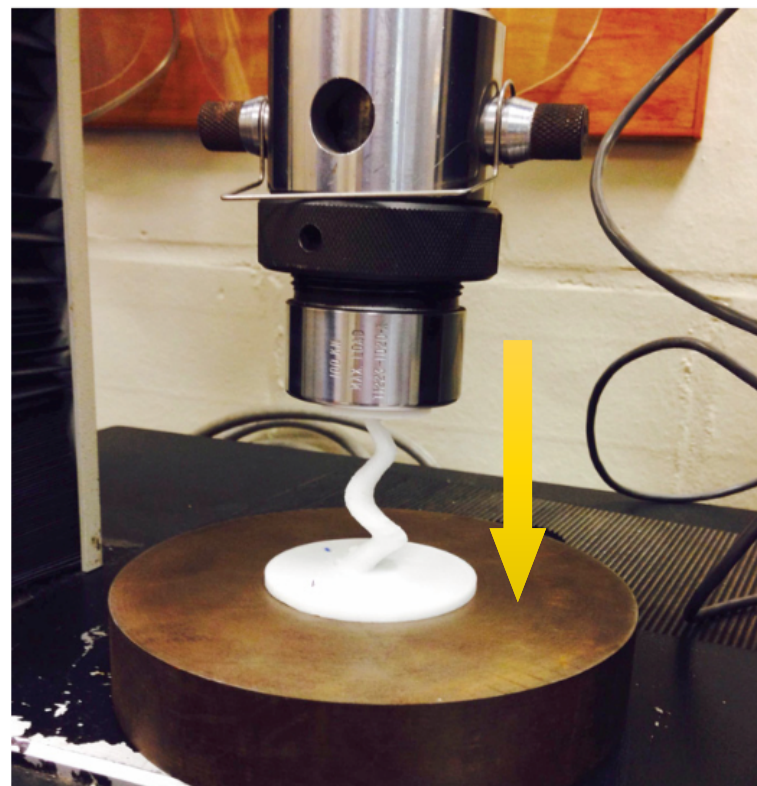


# Gallery of Results – 200% More Force





# Physical Validation

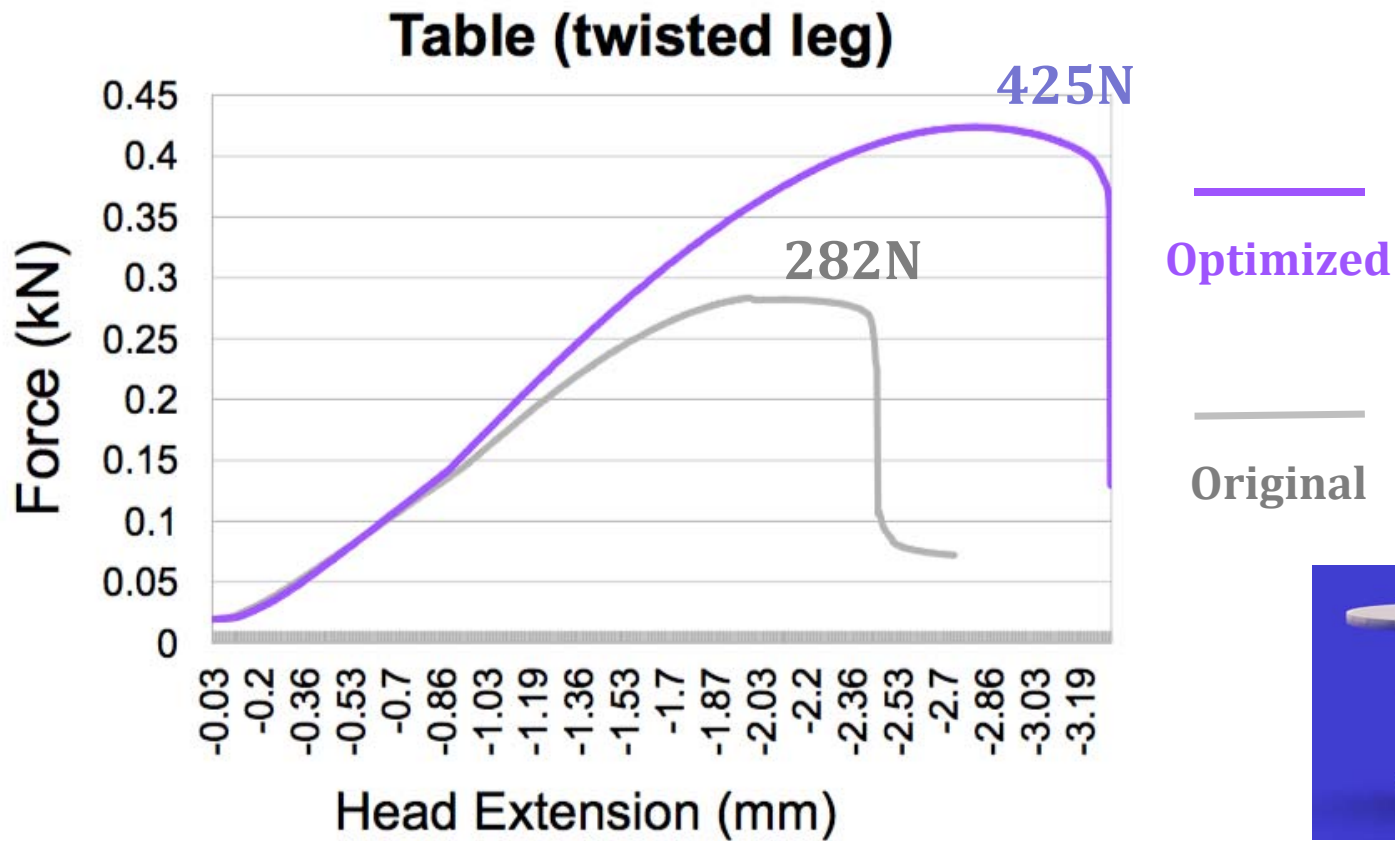


# Physical Validation

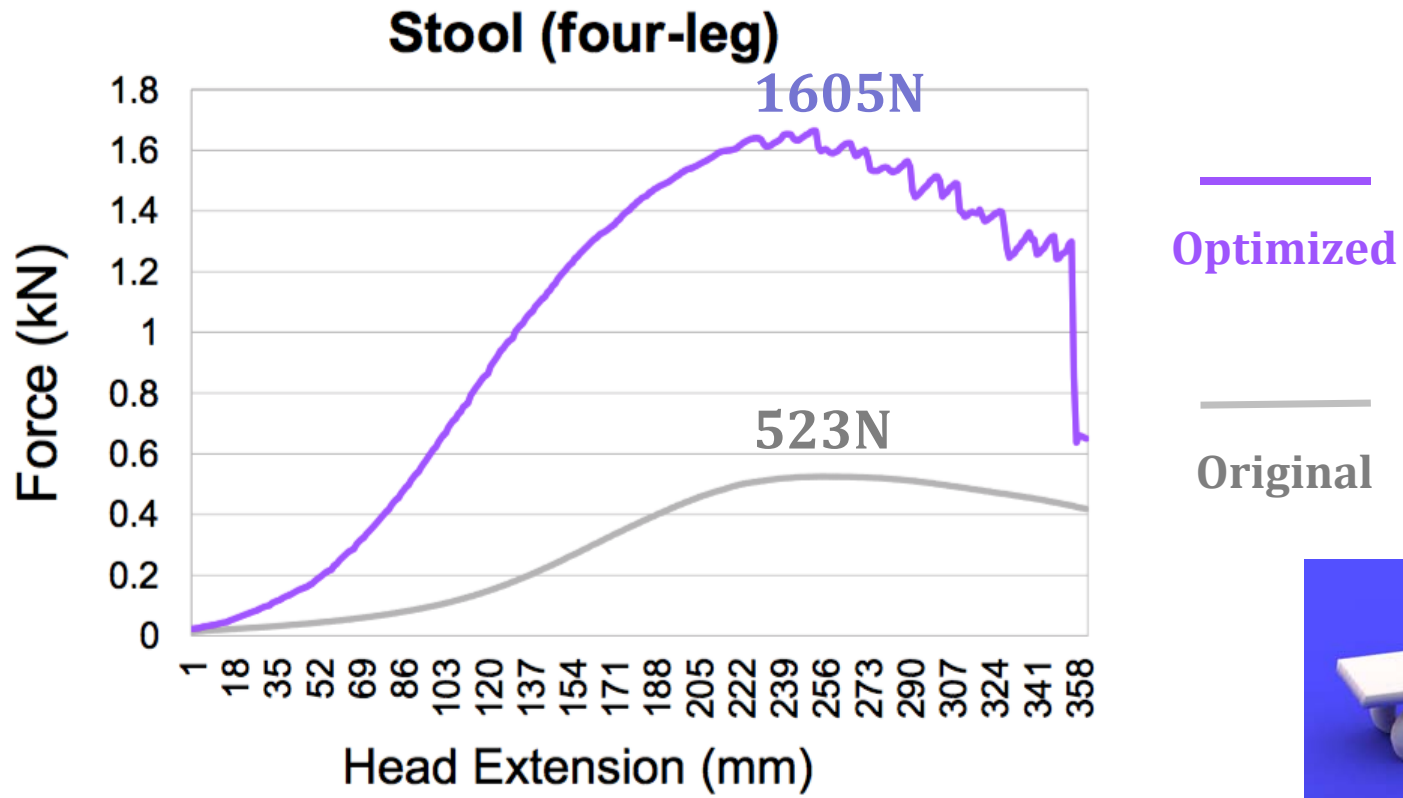
- Optimized shapes withstand 100% more force.
- PLA material, 100% infill, 100 micron resolution
- Equalize volume for fair comparisons.



# Physical Validation



# Physical Validation



# Conclusion

- An algorithm to directly optimize a 3D mesh to make it withstand specified external force.
- Integrates optimization and physics simulation in a unified framework.
- Derivations of analytic gradient and Hessian of the objective function.
- Applications to printable object design.



# Limitations and Future Work

- Performance, convergence speed
- Tetrahedralization quality
- Incorporating higher-order Laplacians [BS08]
- Applications to other design goals, such as improving aerodynamic properties of shapes.



# Acknowledgement

- PG reviewers
- NSF grants CHS-1422441, CHS-1617333, and IIS-1423082.

## Direct Shape Optimization for Strengthening 3D Printable Objects

