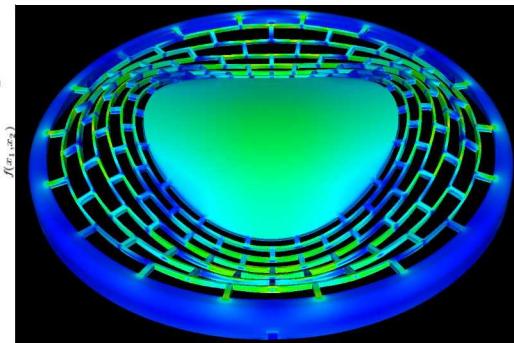
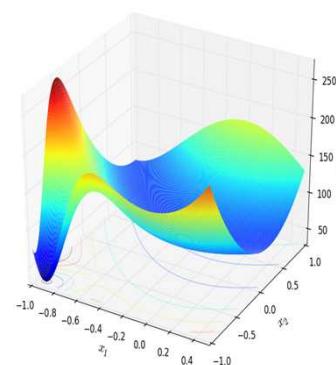


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# A Second-Order Accurate Mathematical Optimization Algorithm Based on Gradient Information

Miguel A. Aguiló

July 29, 2015

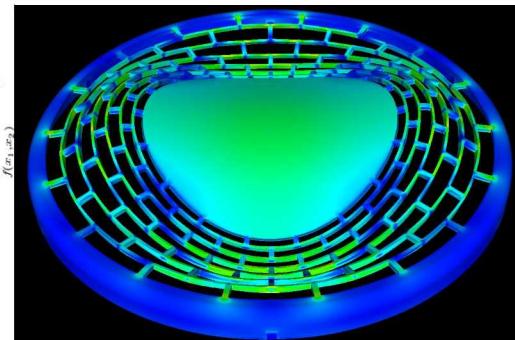
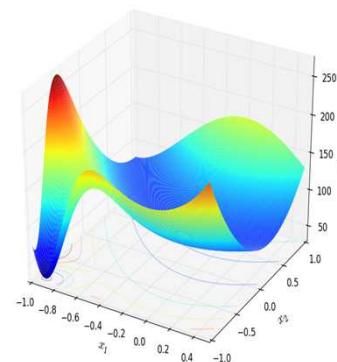


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# Agenda

- Motivation
- Proposed Methodology
- Optimization Algorithm
  - Pseudocode
  - Hessian Information Approximation
- Examples
- Future Research Directions
  - Conclusions

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# Motivation

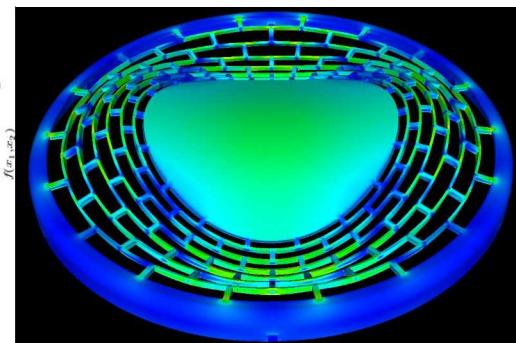
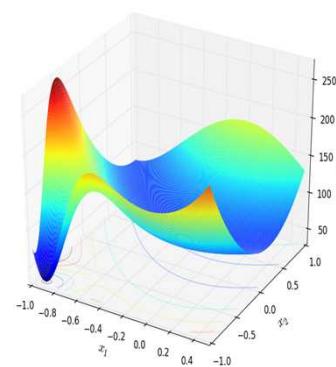
# Motivation

- **Optimization Algorithmic Research**
  - Reduce computational expense of topology optimization problems
  - Minimize tuning parameter
  - Investigate effectiveness of multiple optimization algorithms in the context of topology optimization
- **Questions**
  - Can second-order information reduce computational expense of topology optimization problems?
  - Can we minimize the computational cost associated with computing second-order information?

# Proposed Methodology

- **Objective:**
  - Develop a general optimization algorithm that enables the computation of *second-order* (i.e. *Hessian*) information from the available *first-order* (i.e. *gradient*) information
- **Requirements:**
  - Second-order derivative operators are not required
  - Computational cost does not increase with the number of design variables
  - Computational cost is the same as solving the topology optimization problem with just gradient information

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# Optimization Algorithm



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# Algorithm Pseudocode

## Outer Loop

Set initial  $x_{i=0}$

Compute  $J(x_{i=0})$

while(not converge)

{

  Solve Sub-problem

  Check convergence

$i = i + 1$

}

## Legend

$x$  = Control

$J$  = Objective function

$G$  = Gradient

$H$  = Hessian

$d$  = Trial step

## Line Search Sub-problem

{

  Compute  $G(x_i)$

  Apply filter  $F(G_i)$

  Solve  $H(x_i)d_i = -G(x_i)$

  Line search scaling  $s_i$

$x_{i+1} = x_i + s_i d_i$

}

## Trust Region Sub-problem

while(TR criteria met)

{

  Compute  $G(x_i)$

  Apply filter  $F(G_i)$

  Solve  $H(x_i)\hat{d}_i = -G(x_i)$ ,  $\hat{d}_i = s_i d_i$

  Check TR sub-problem convergence

    -  $x_{i+1} = x_i + \hat{d}_i$

}

# Hessian Approximation: Finite Differences (FDiff)

## Finite Difference

Given  $h > 0$ , the  $d^{\text{th}}$  order derivative of an univariate function  $f(x)$  can be approximated by finite differences as follows:

**Forward**  $\frac{h^d}{d!} F^d(x) + O(h^{d+p}) = \sum_{i=i_{\min}}^{i_{\max}} C_i F(x + ih)$

**Backward**  $(-1)^d \frac{h^d}{d!} F^d(x) + O(h^{d+p}) = (-1)^d \sum_{i=i_{\min}}^{i_{\max}} C_i F(x + ih)$

**Centered** = *Forward - Backward*

where  $p > 0$  denotes the order of accuracy,  $i_{\min}$  and  $i_{\max}$  denote some choice of extreme indices , and coefficients  $C_i$

# Cont: Finite Difference

## Example

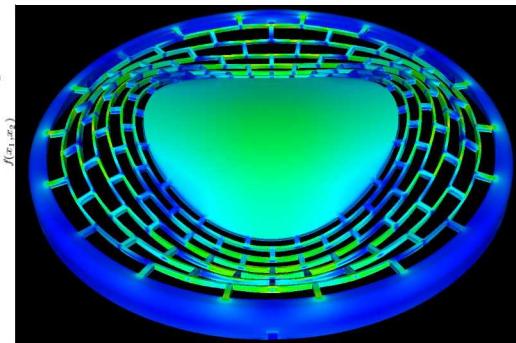
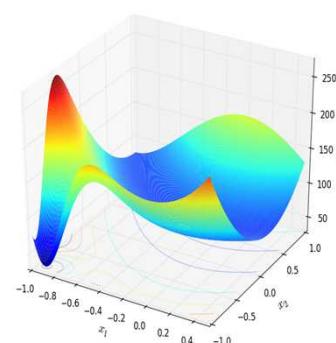
Given gradient operator  $G(x_i)$ , approximate the application of the trial step to the Hessian operator, i.e.  $H_i(x_i)d_i$  through finite differences as follow

**Backward** 
$$G'(x_i) = \frac{3G(x_i) - 4G(x_i - hd_i) + G(x_i - 2hd_i)}{2h} + O(h^2)$$

## Key Features

- Full Hessian operator is not assemble
  - Interested in the application of the trial step to the Hessian operator
- Each evaluation is independent and thus parallelizable
  - Task parallelism can be effectively employ to expedite computation

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## Examples



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# Case Study: Compliance Minimization

## Optimality Criteria (OC) Formulation

$$\begin{aligned} \min \quad & \frac{1}{2} \mathbf{u}^T (z^p \mathbf{K}) \mathbf{u} + R(z) \\ \text{s.t.} \quad & (z^p \mathbf{K}) \mathbf{u} - f = 0 \\ & \frac{V(z)}{V_o} \leq \gamma \end{aligned}$$

## Nonlinear Programming Solver's Formulation

$$\begin{aligned} \min \quad & \frac{1}{2} \mathbf{u}^T (z^p \mathbf{K}) \mathbf{u} + R(z) + \frac{1}{2} (V(z) - \gamma V_o)^2 \\ \text{s.t.} \quad & (z^p \mathbf{K}) \mathbf{u} - f = 0 \end{aligned}$$

## Test Problems



Mitchell

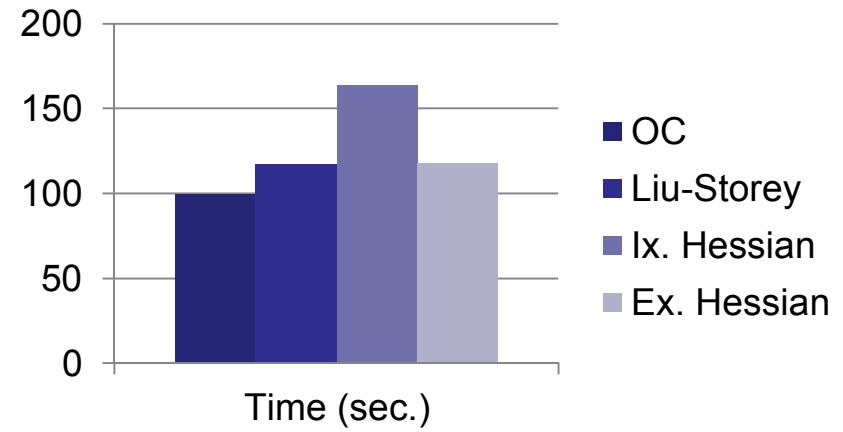
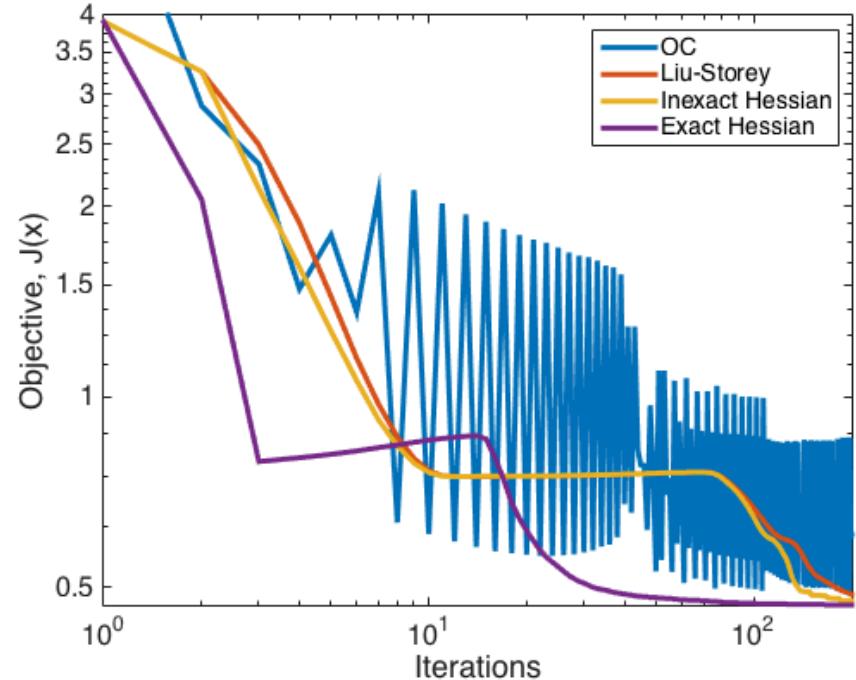
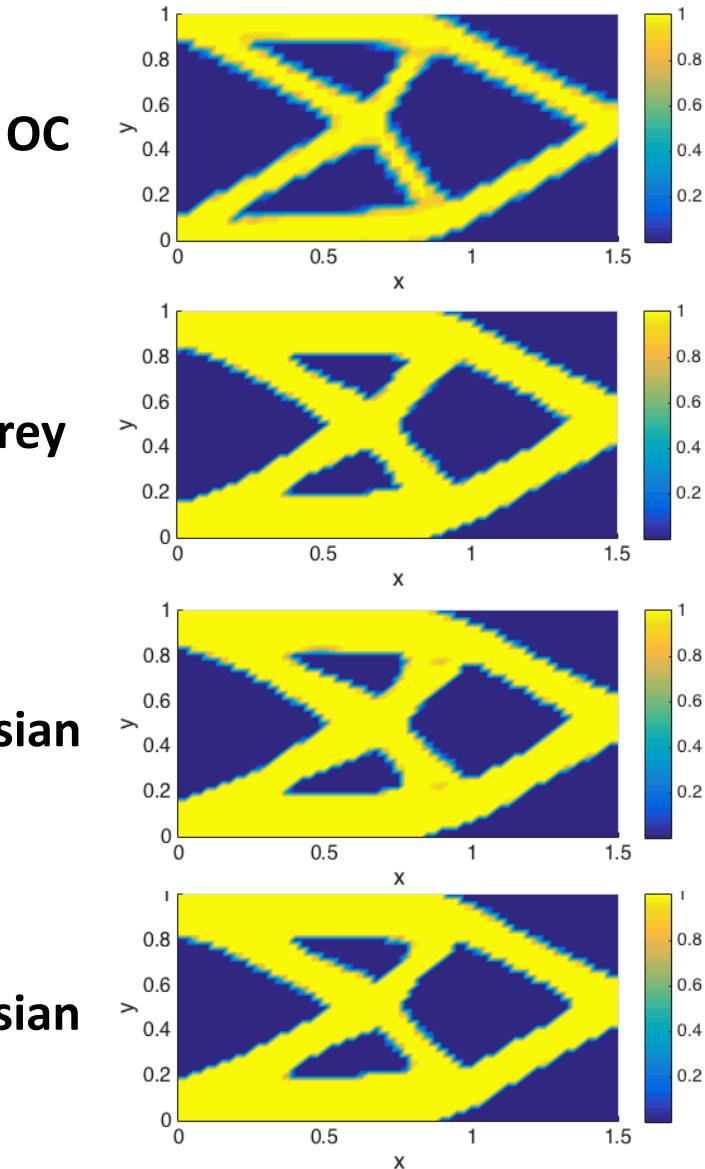


Cantilever

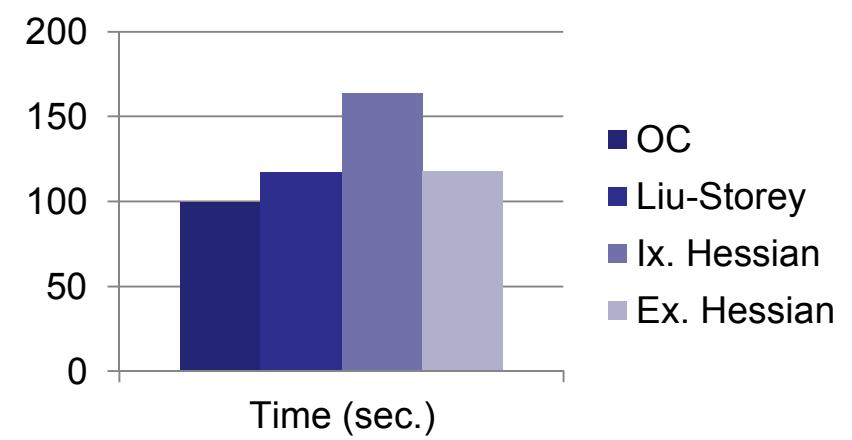
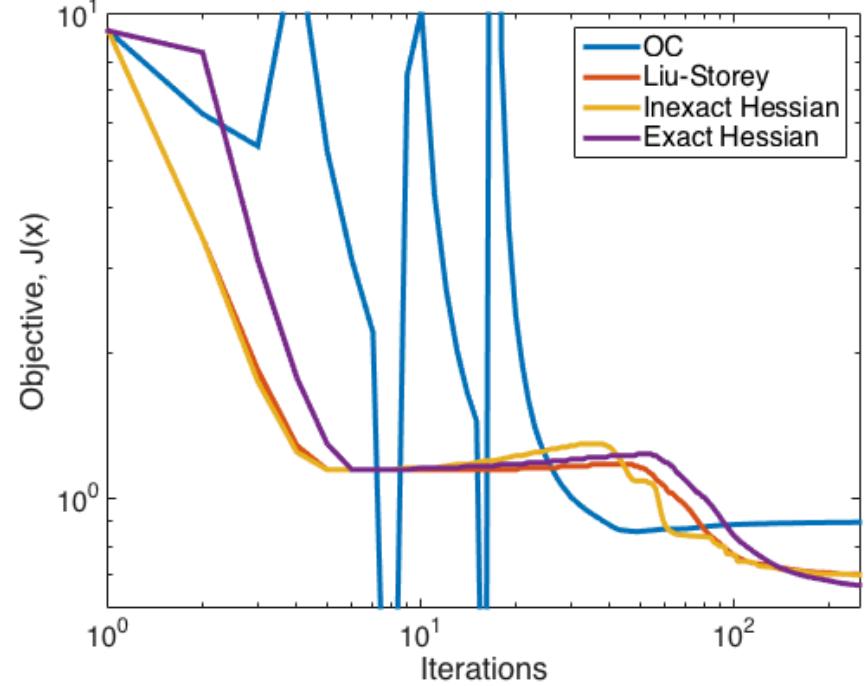
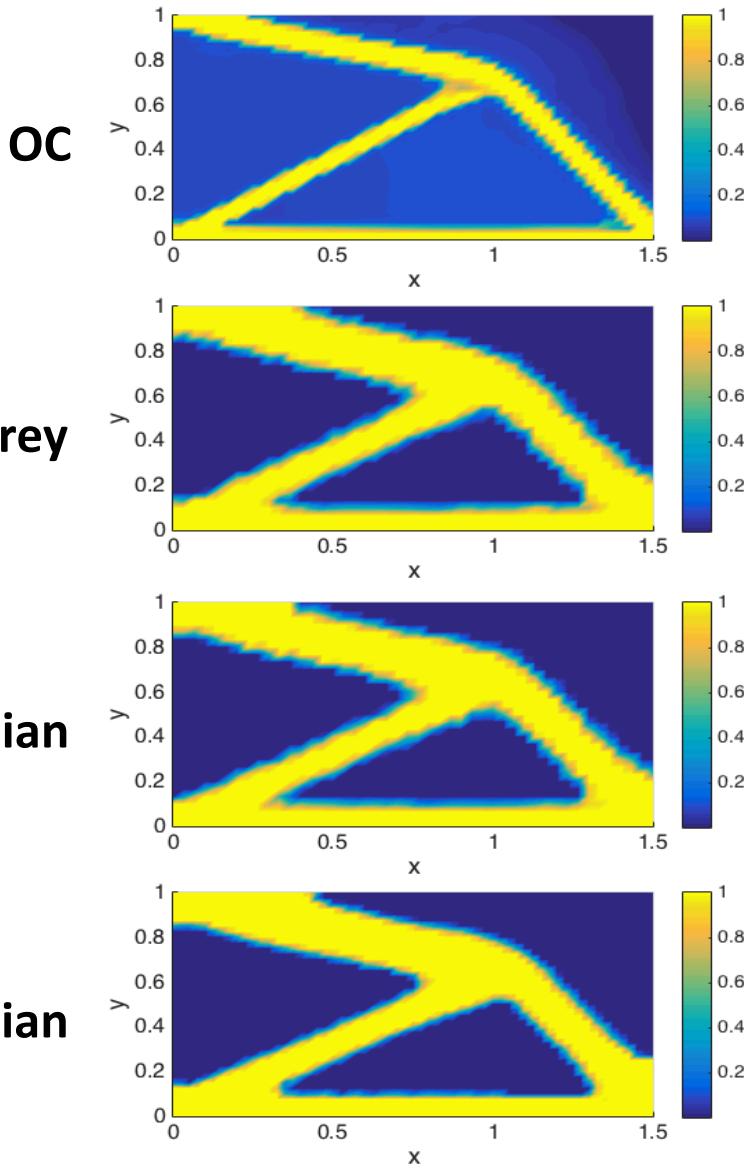


MBB

# Example 1: 60x40 FE Mesh

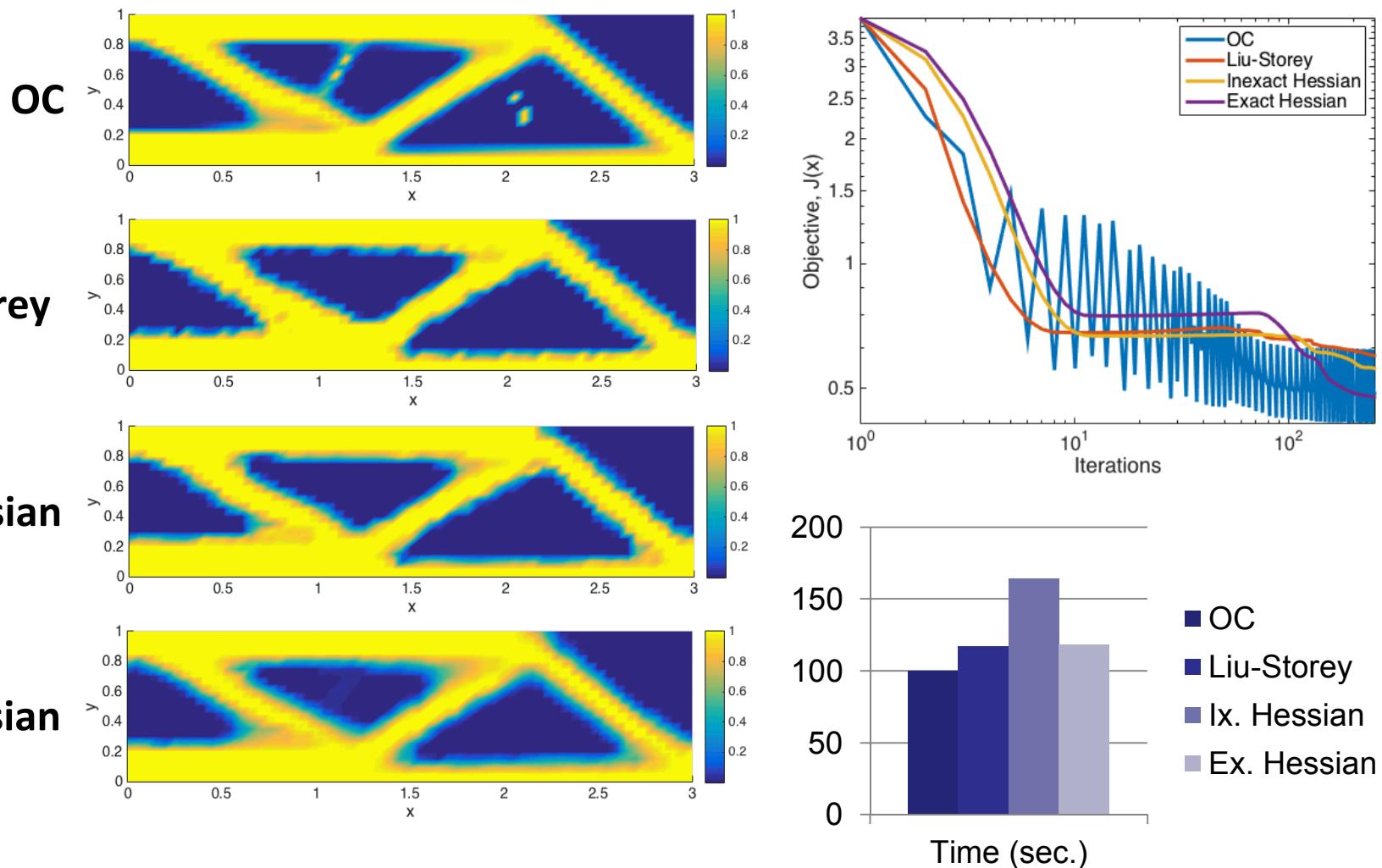


# Example 2: 60x40 FE Mesh

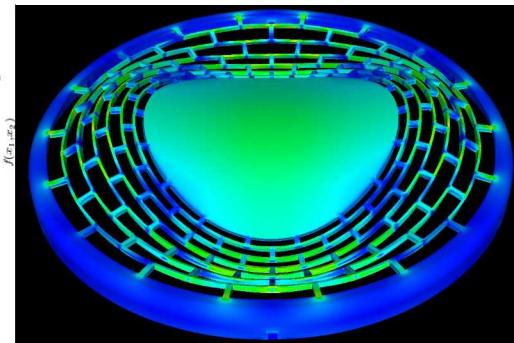
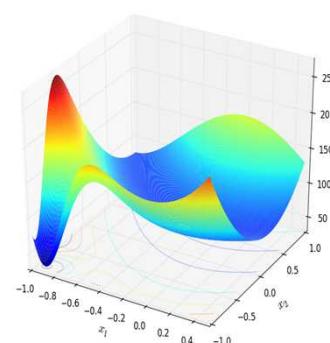


- OC
- Liu-Storey
- Ix. Hessian
- Ex. Hessian

# Example 3 (60x20)



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## Future Research Directions

# Final Remarks

- **Additional tasks**
  - Compare performance against MMA and GCMMA algorithms
  - Exercise formulation based on filters and stress minimization problems
- **Algorithmic Improvements**
  - Research in primal-dual interior-point methods
  - Explore task parallelism of solves needed for inexact Hessian computation
  - Reduced order model based optimization algorithm
    - Allow optimization algorithm to automatically generate and update reduced order model for physics