

# Direct numerical simulations in solid mechanics for quantifying the effects of microstructure and material-model form error on macroscale quantities of interest

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# Outline

- research questions
- direct numerical simulations in solid mechanics
- example: elastic structure with microstructure
- example: plastic response
- applications to additive manufacturing
- multiscale modeling using geometric multigrid



# Research questions (why DNS?)

- What is “material variability”?
- What is the error in homogenization theory?
- What is the error induced when the assumption of scale-separation no longer holds?
- Can we find evidence of **surface-effects**?
- Can we find evidence of strain-gradient effects or **nonlocality** in the mean-field response?
- How to include known spatial variations in microstructure in our macroscale simulations, e.g. arising from manufacturing processes?
- How does material variability impact engineering quantities of interest?

# Direct numerical simulations

- Key postulate: we have a fine scale representation/model that is predictive (e.g. microstructural model, crystal-plasticity model)
- Perform direct numerical simulations (DNS) of macroscopic boundary-value problems with microstructure and compare with the solution from the homogenized PDE.
- Identify any evidence of higher-order effects (gradient or surface effects).
- Can currently model  $\sim 500,000$  grains in a macroscale structure.
- Goal is  $\sim 100$  M grains.

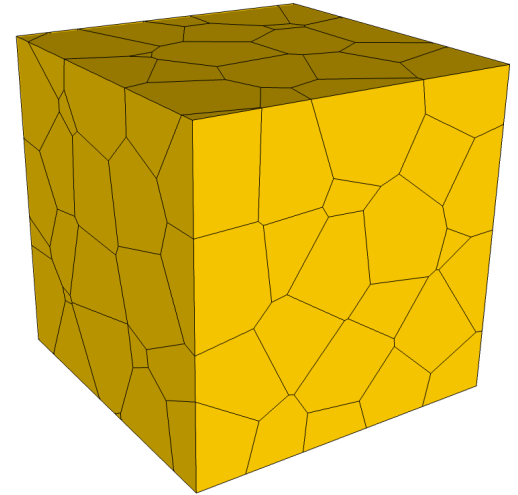
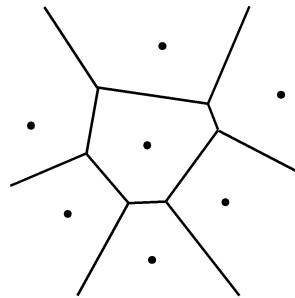
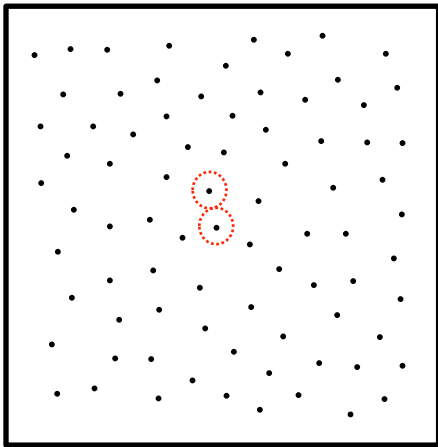
# DNS Approach

- Use voxelation approach to mesh grains
- Use macroscale hexahedral mesh as “overlay” grid
- Use idealized Voronoi microstructure (for now)
- Use Maximal Poisson Sampling (MPS) to seed Voronoi microstructure (results in equiaxed grain morphology)
- No texture (for now)

## Advantage of voxelation approach

- Easy meshing of microstructure
- Only need implicit representation of microstructure
- Can robustly generate many microstructural realizations
- Robust under large deformation
- Experimental data is typically pixelated anyway

# Voronoi Microstructure from MPS Seeding

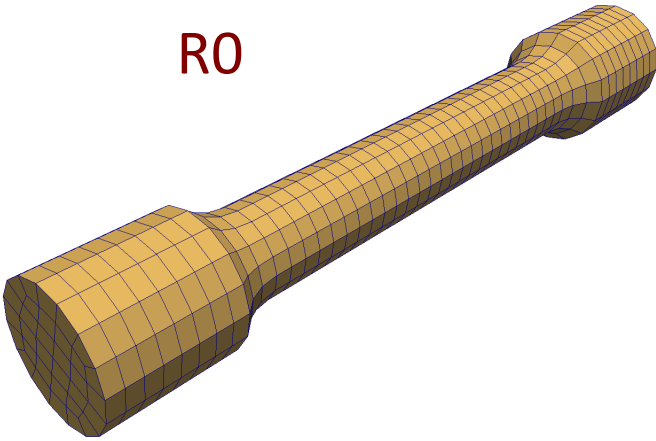


## Maximal Poisson Sampling

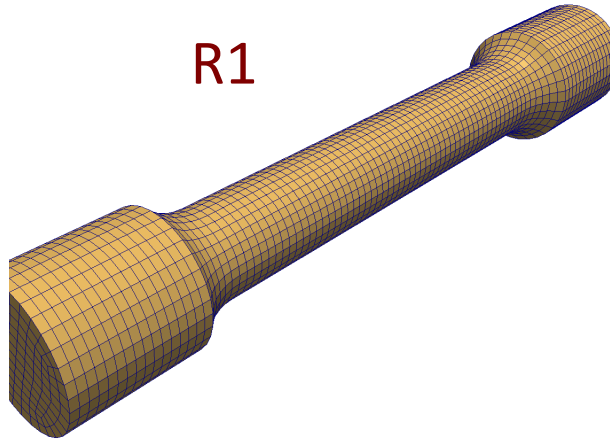
- constraint on minimum distance
- seed until 'max' packing
- Ebeida/Mitchell Algorithm (1400)

# Hierarchy of hexahedral meshes

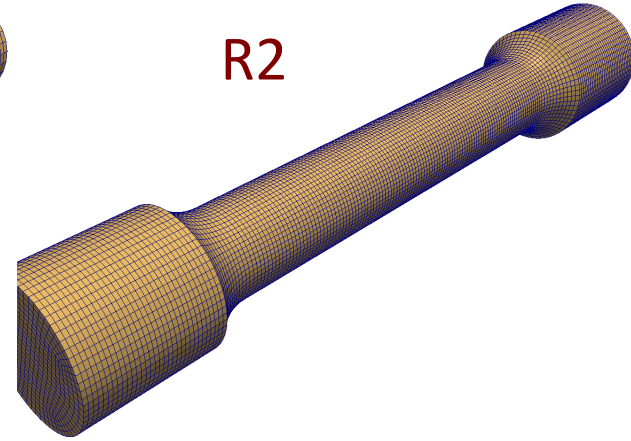
R0



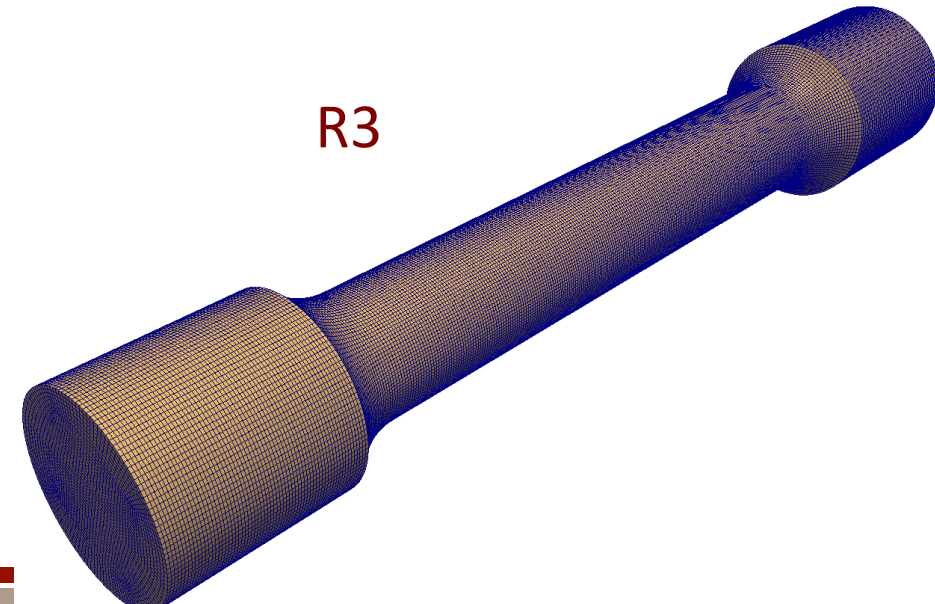
R1



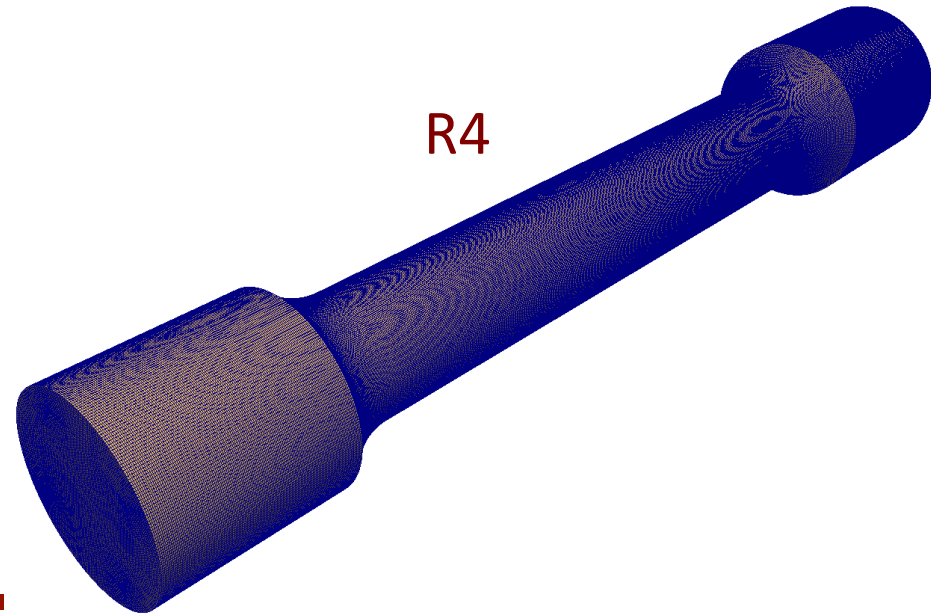
R2



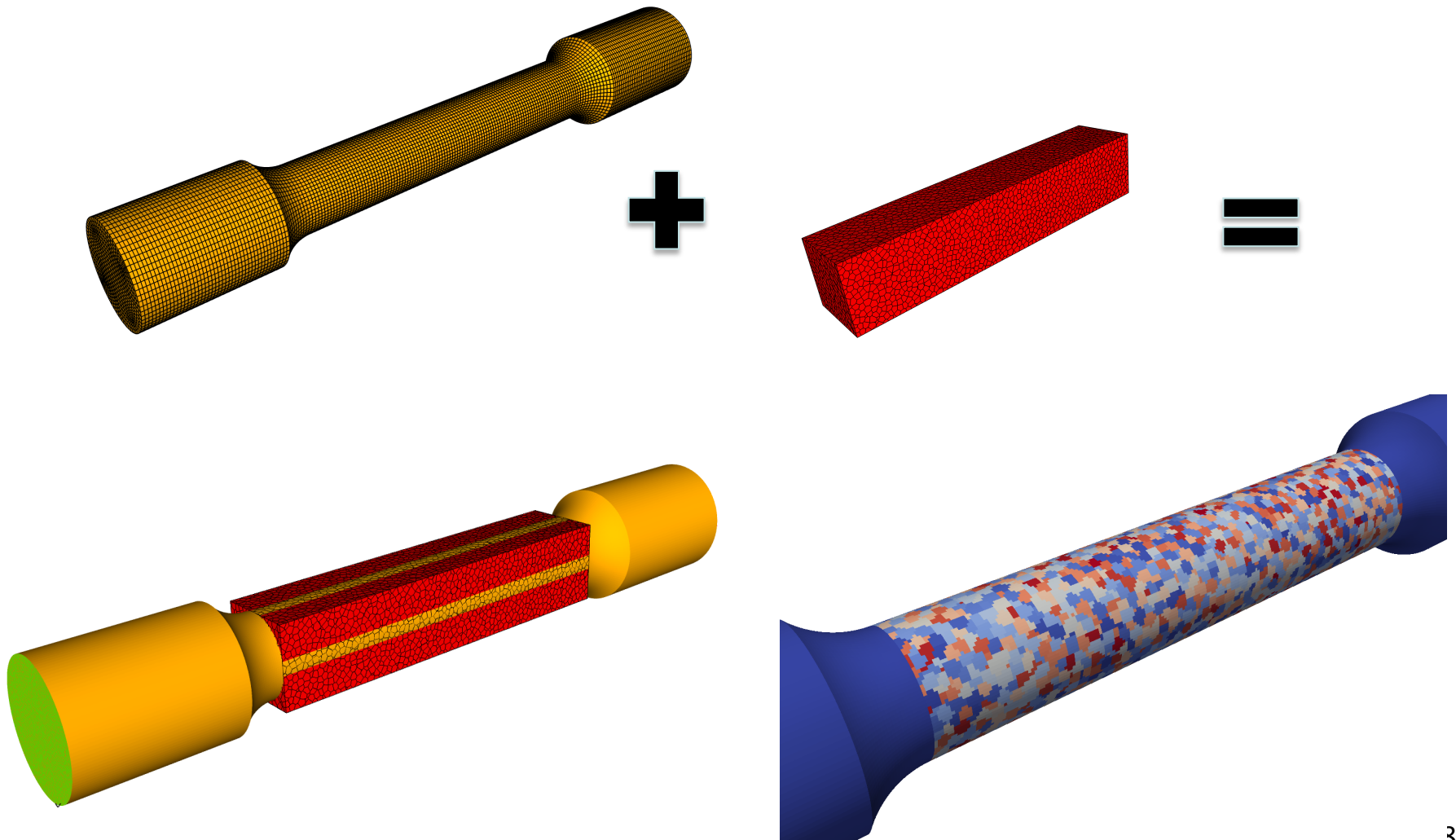
R3



R4



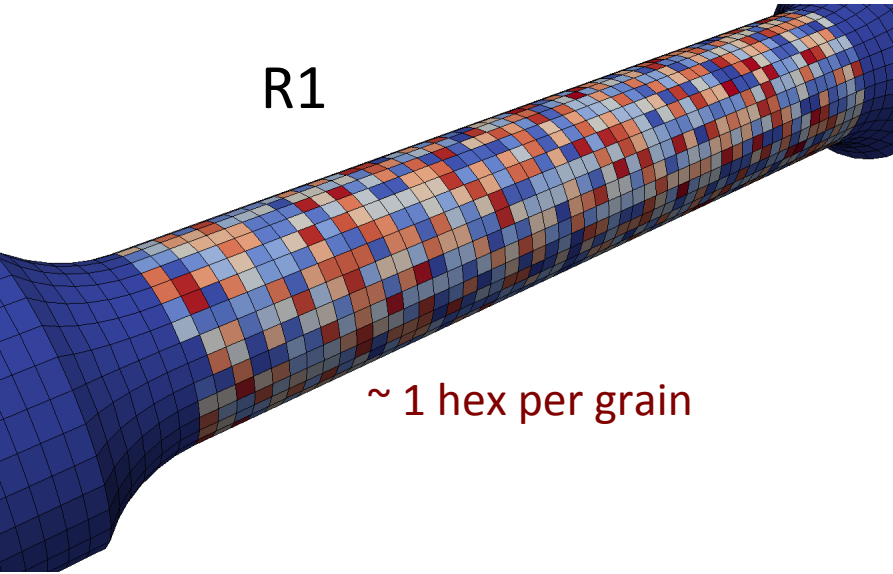
# Microstructural overlay of hex mesh



# Voronoi overlay of hierarchy of hex meshes

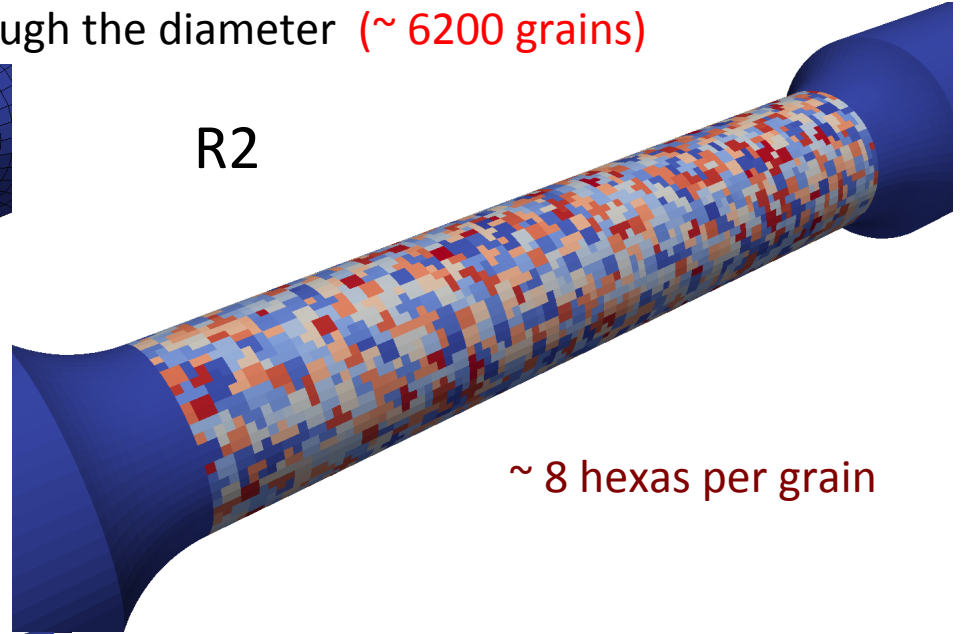
One grain realization with  $\sim 12$  grains through the diameter ( $\sim 6200$  grains)

R1



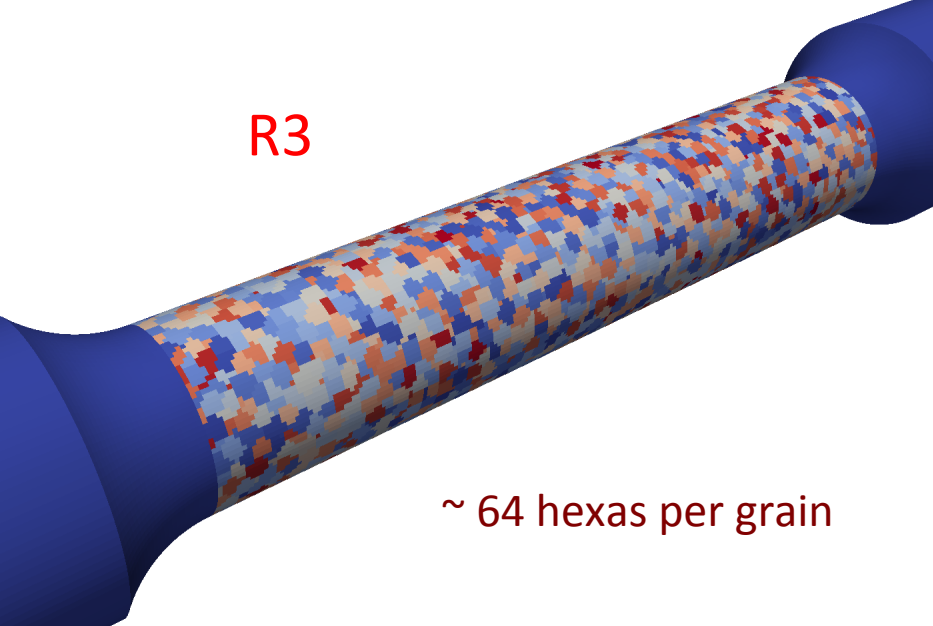
$\sim 1$  hex per grain

R2



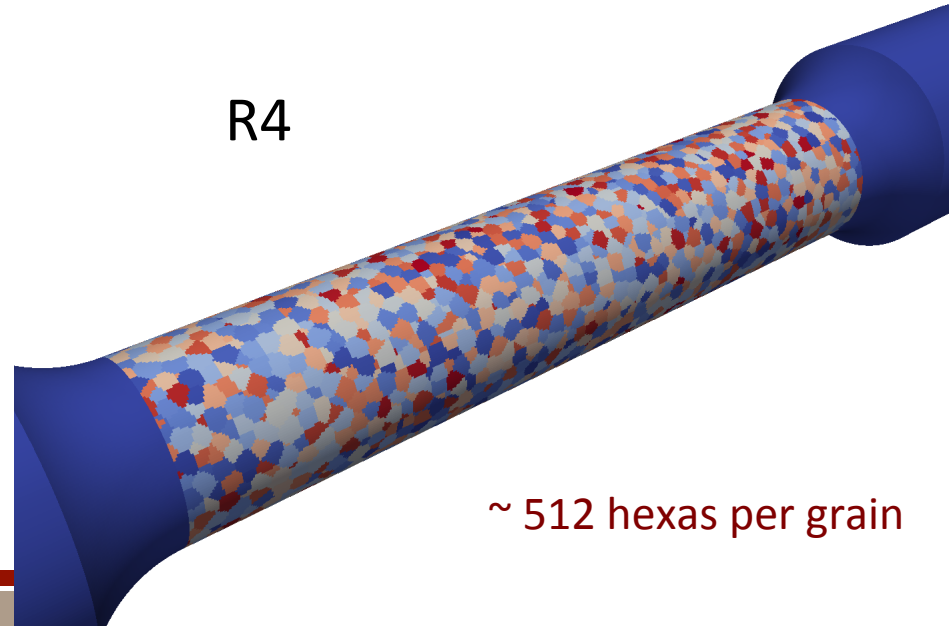
$\sim 8$  hexas per grain

R3



$\sim 64$  hexas per grain

R4



$\sim 512$  hexas per grain

# Stainless steel 304L single crystal elasticity constants

(Ledbetter, 1984)

single crystal elastic constants (cubic symmetry)

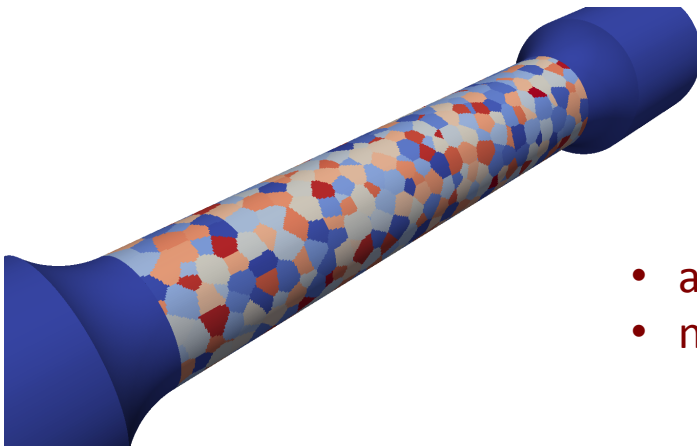
$$C_{11} = 204.6 \text{ GPa}$$

$$C_{12} = 137.7 \text{ GPa}$$

$$C_{44} = 126.2 \text{ GPa}$$

anisotropy ratio,

$$A = \frac{2C_{12}}{C_{11} - C_{44}} = 3.5$$



- assume random crystallographic orientations
- no correlation between grains (no texture)



# RPI crystal plasticity model

plastic velocity gradient:  $L^p = \sum_{\alpha=1}^N \dot{\gamma}^{\alpha} P^{\alpha}$  (sum over slip systems)

Schmid tensor:  $P^{\alpha} = m^{\alpha} \otimes n^{\alpha}$

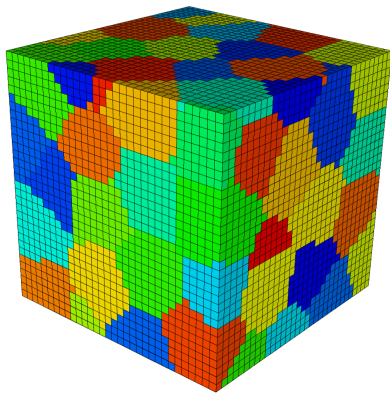
slip system slip rates:  $\dot{\gamma}^{\alpha} = \dot{\gamma}_o \frac{\tau^{\alpha}}{g^{\alpha}} \left| \frac{\tau^{\alpha}}{g^{\alpha}} \right|^{1/m-1}$

slip system hardening:  $g = g_o + (g_{so} - g_o) \left[ 1 - \exp \left( -\frac{G_o}{g_{so} - g_o} \gamma \right) \right]$

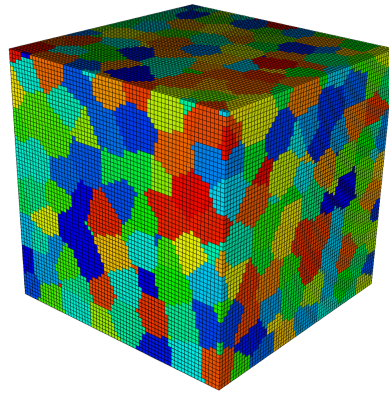
$$\gamma = \sum_{s=1}^N |\gamma^s|$$

# How to get homogenized properties?

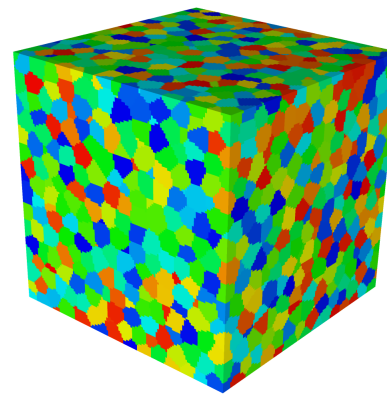
Convergence of stochastic volume elements (SVE) to representative volume element (RVE)



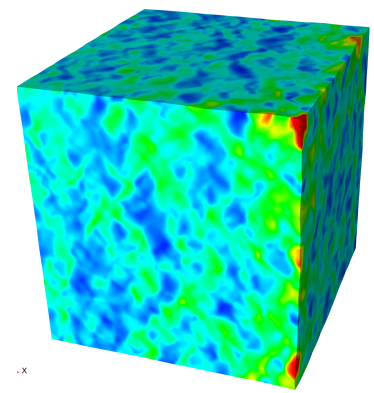
$\sim 4^3$  grains



$\sim 8^3$  grains



$\sim 16^3$  grains



plastic response

# Convergence to effective isotropic elastic properties

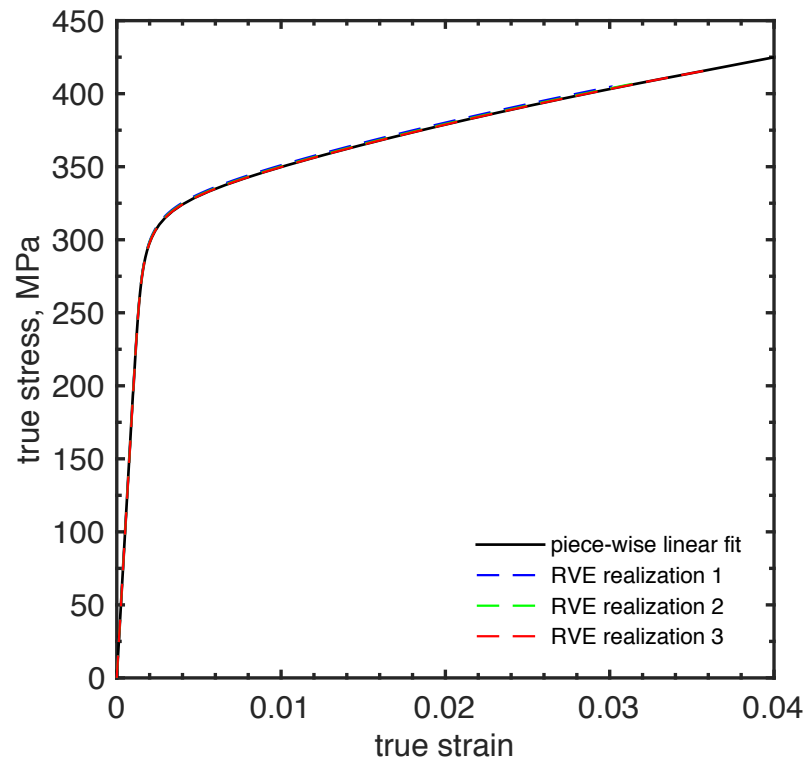
- mean of 100 simulations at each “grain level”
- rational function extrapolation to  $\infty$

number of grains	apparent Young's Modulus (GPa)	apparent Poisson's ratio
$\sim 4^3$ grains	185.2	0.307
$\sim 8^3$ grains	190.5	0.301
$\sim 16^3$ grains	193.9	0.298
$\sim 32^3$ grains	195.7	0.296
$\infty$	197.6	0.294

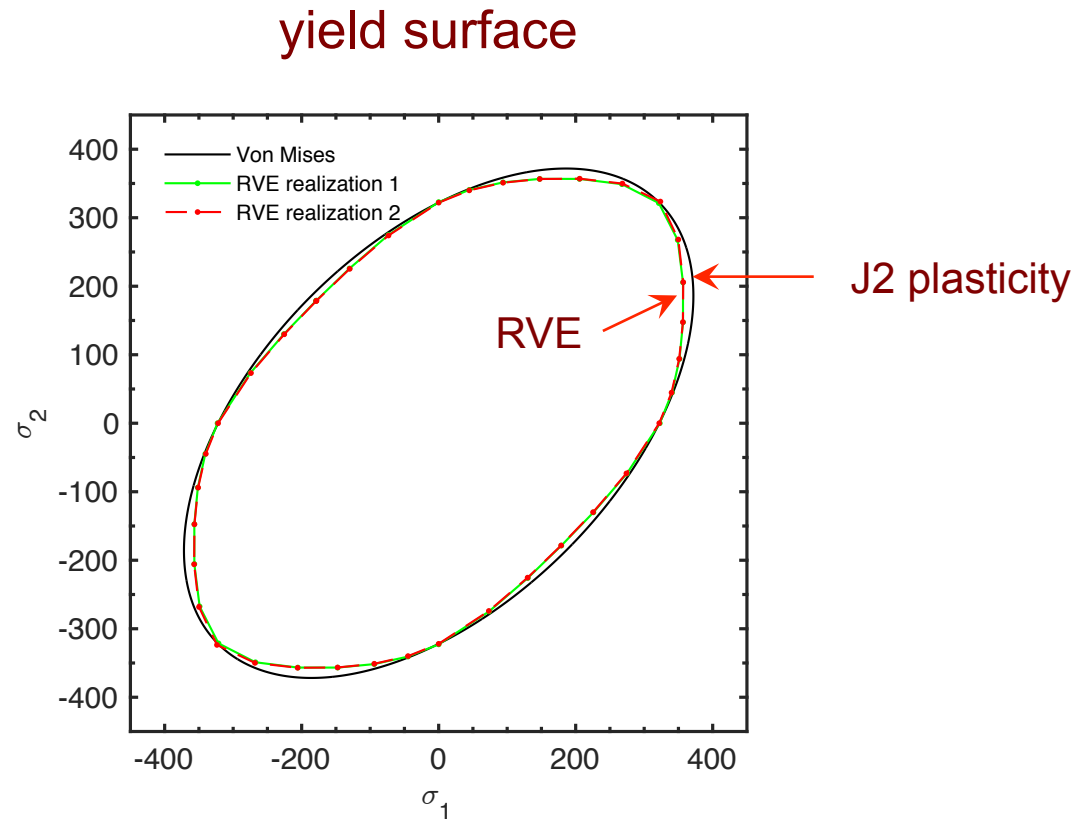
These values will be used as the homogenized, isotropic, elastic properties.

# Effective plasticity model?

- Ideally, would use computational homogenization (FE<sup>2</sup>) for nonlinear homogenization.
- Since this is not available, use a simple piece-wise linear hardening J2-plasticity model. This results, however, in a model-form error.

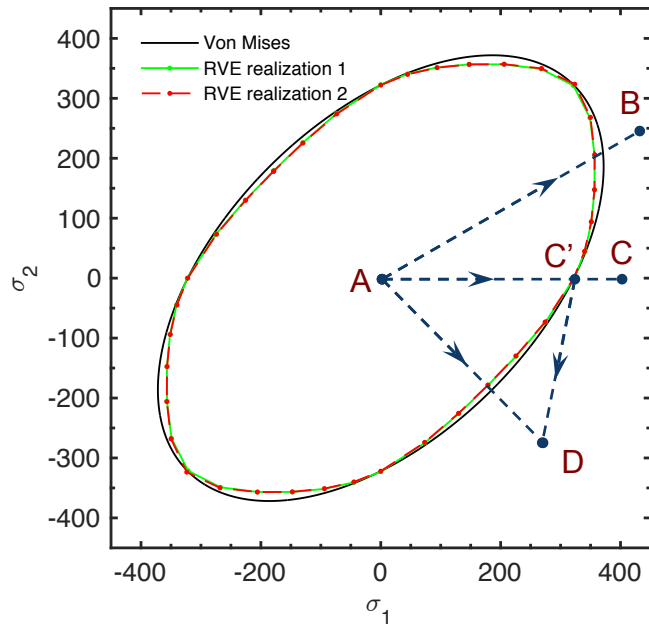


# Model-form error – RVE vs. J2-plasticity



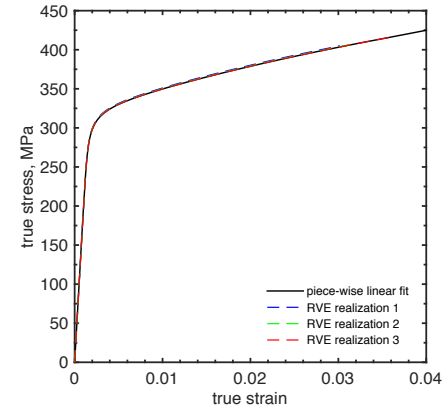
# Model-form error – RVE vs. J2-plasticity

stress paths in  
principal-stress space



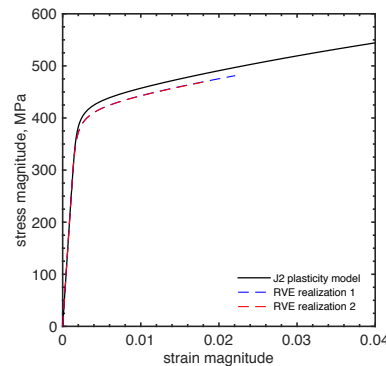
calibration

stress path A – C

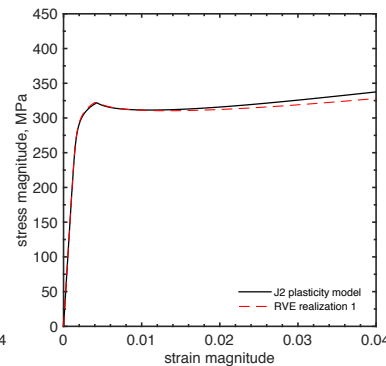


validation

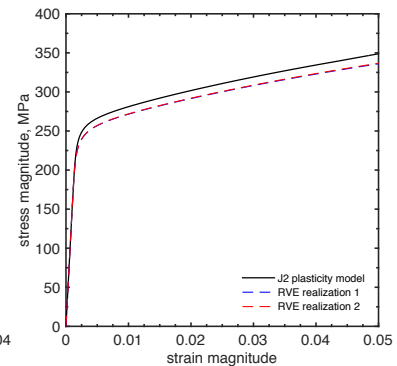
stress path A – B



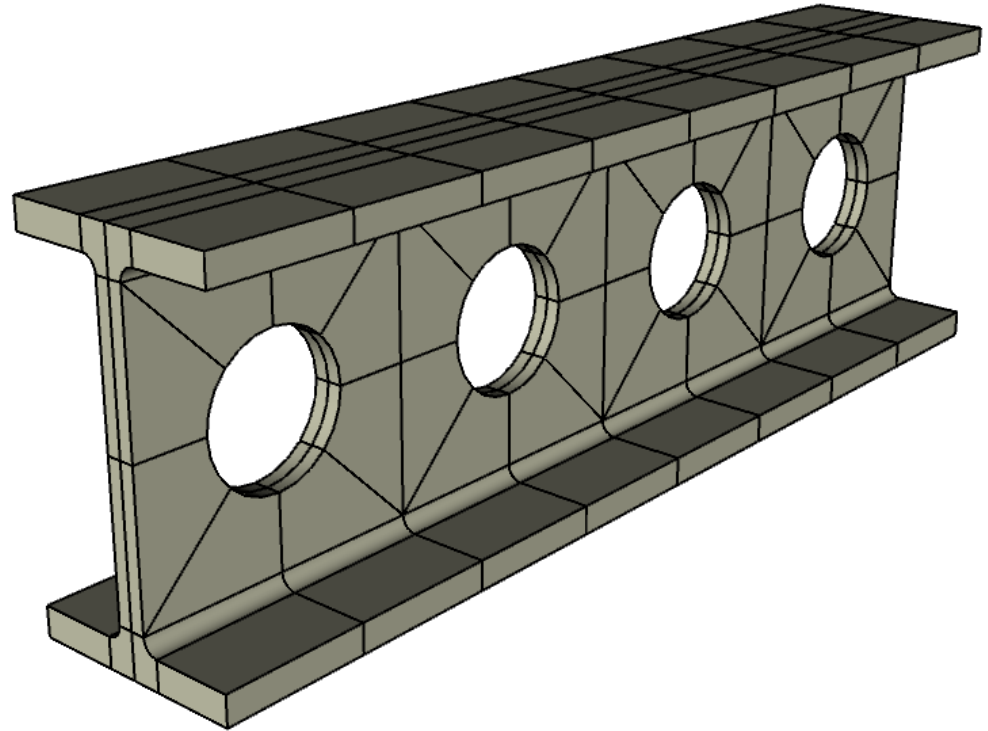
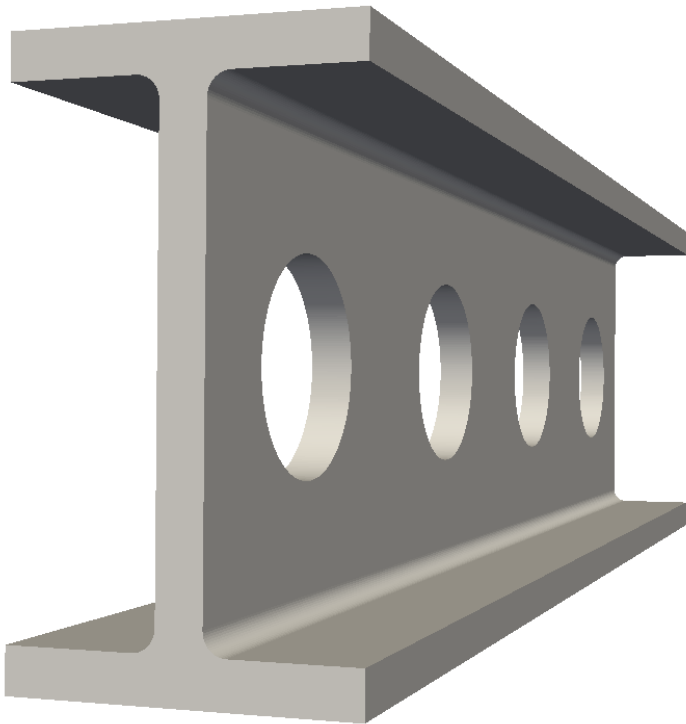
stress path A – C' – D



stress path A – D

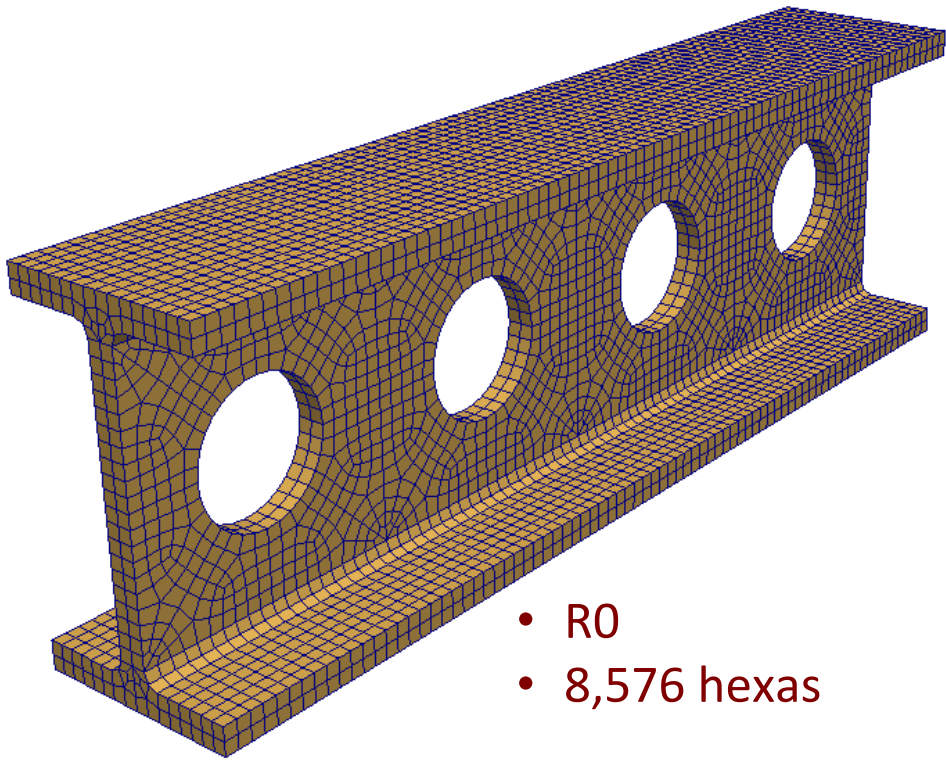


# I-Beam example - elastic

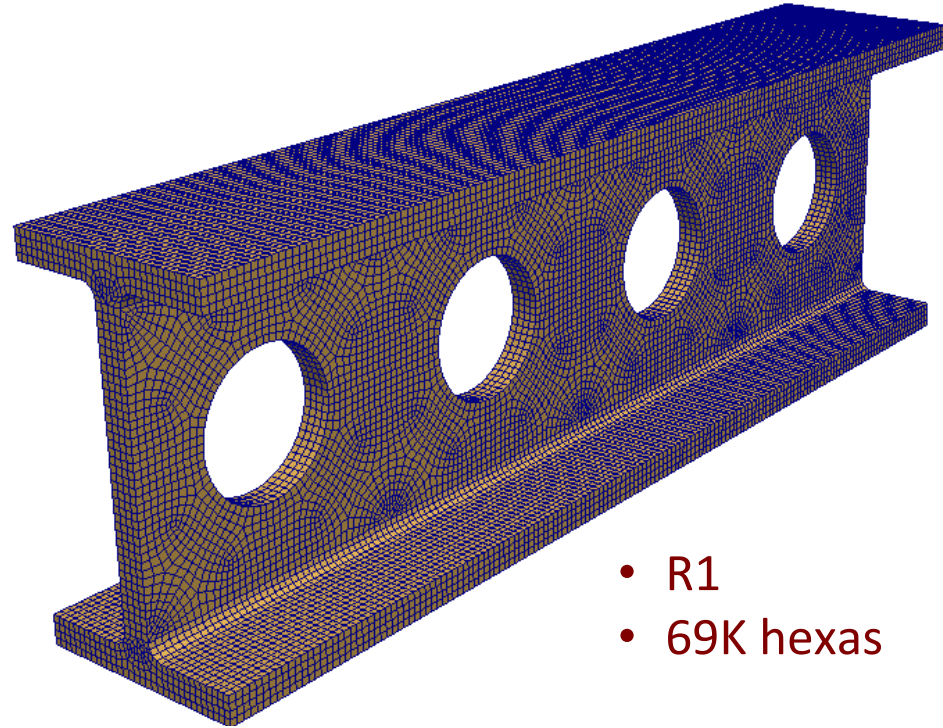


- tension
- bending
- torsion

# Hierarchy of hexahedral meshes



- R0
- 8,576 hexas

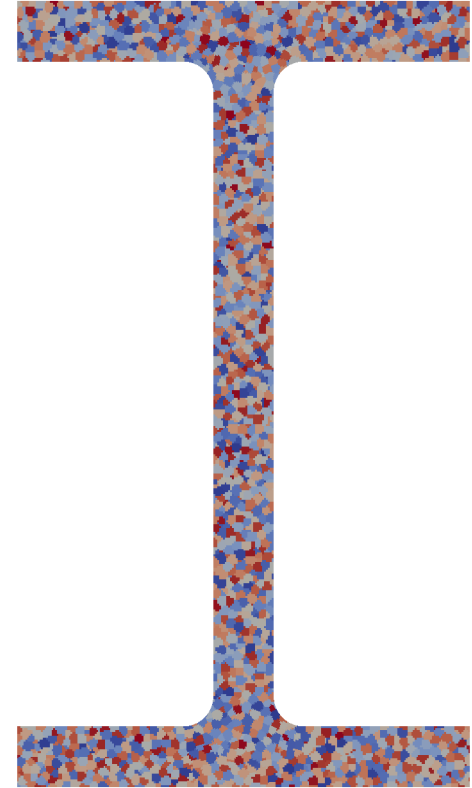
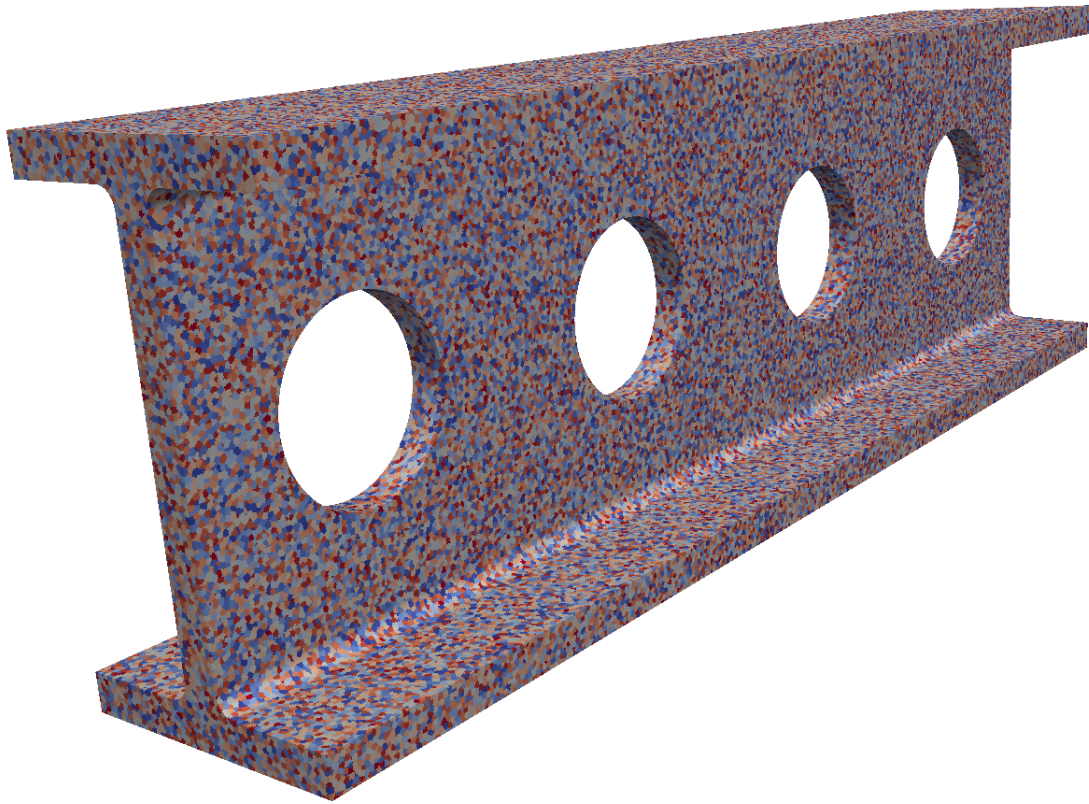


- R1
- 69K hexas

- R2, 549K hexas
- R3, 4.4M hexas
- R4, 35M hexas (~ 2000 cores, FETI solver)
- R5, 280M hexas (~ 20,000 cores, 3-level FETI)

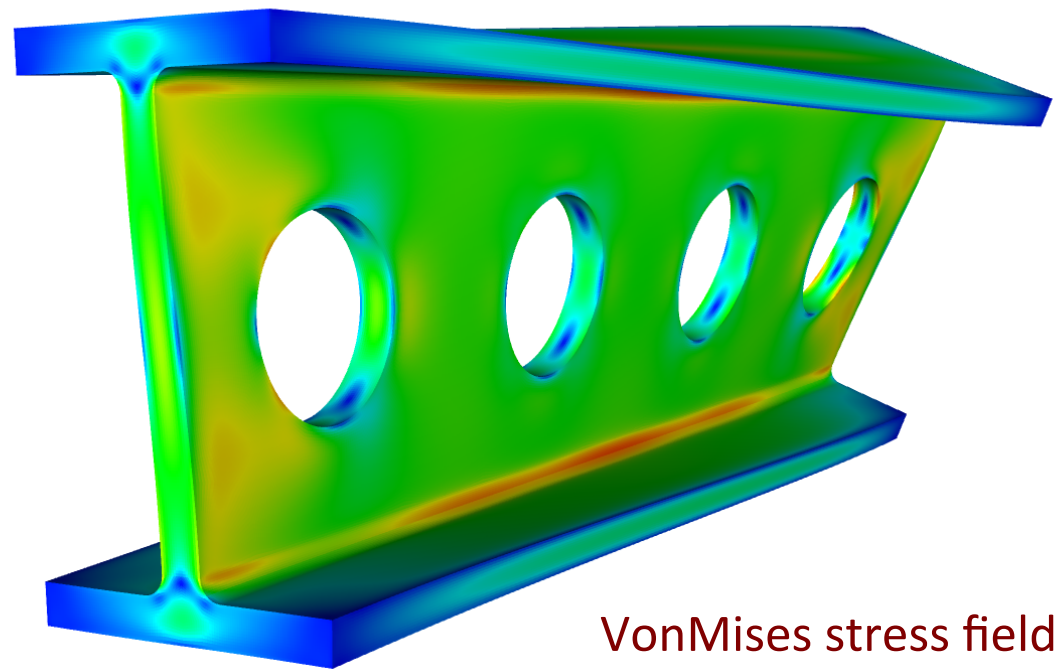


# Thickness/grain ratio = 8

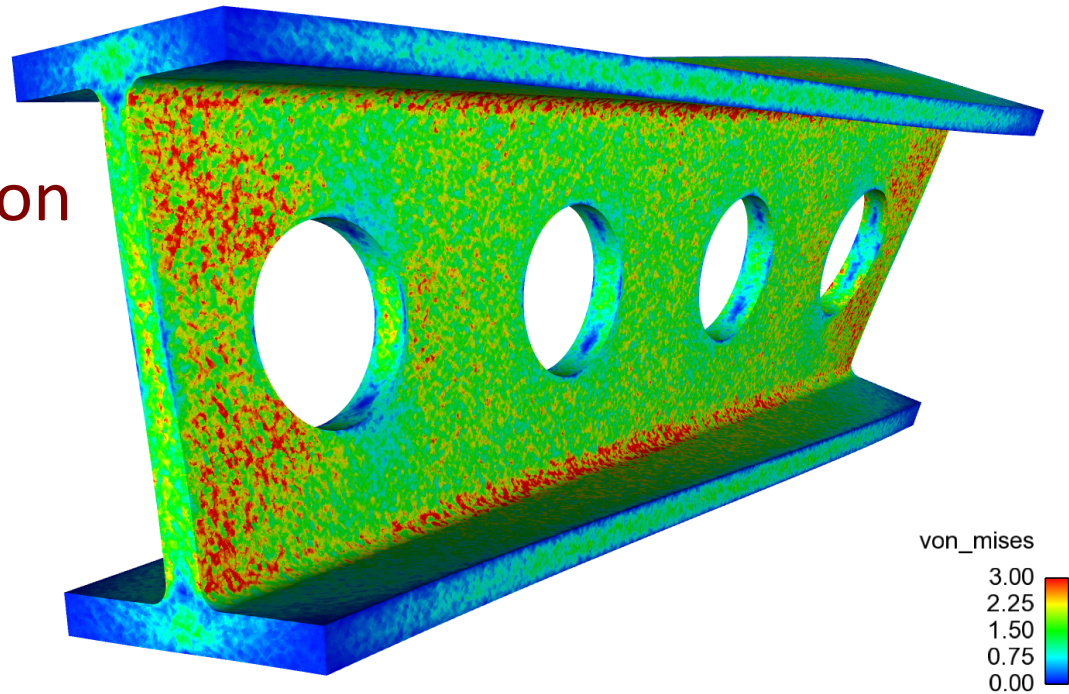


- uniformly random crystal orientations
- ~420,000 grains
- hex mesh overlay = R4 (35M elements)

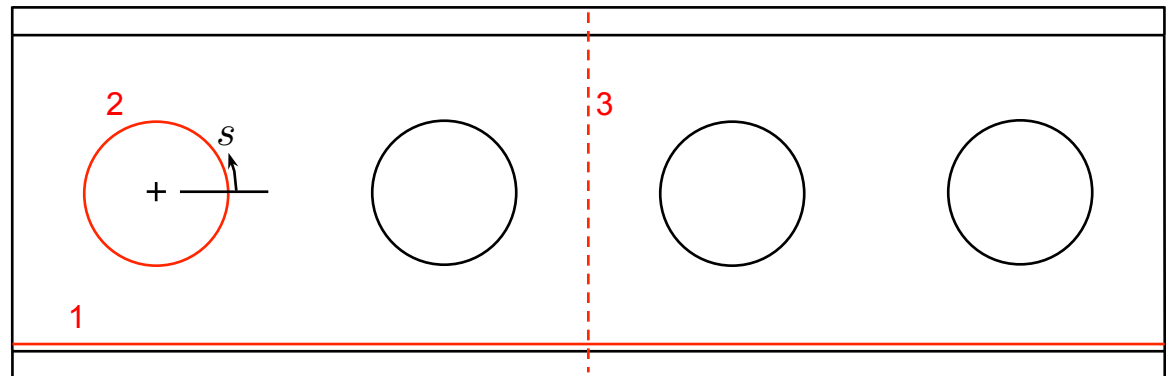
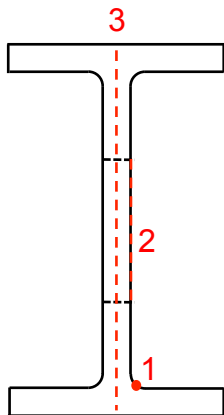
homogenized solution



direct numerical simulation



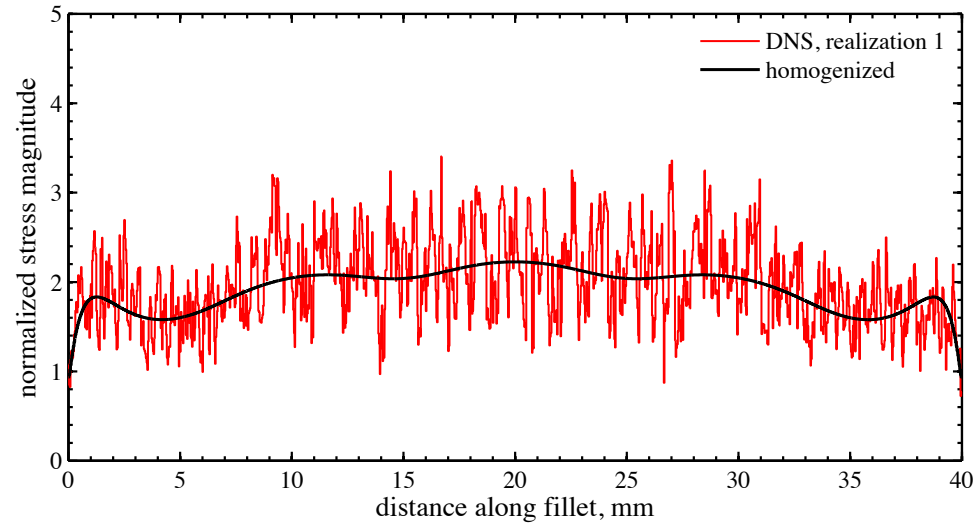
# Stress extraction lines/curves



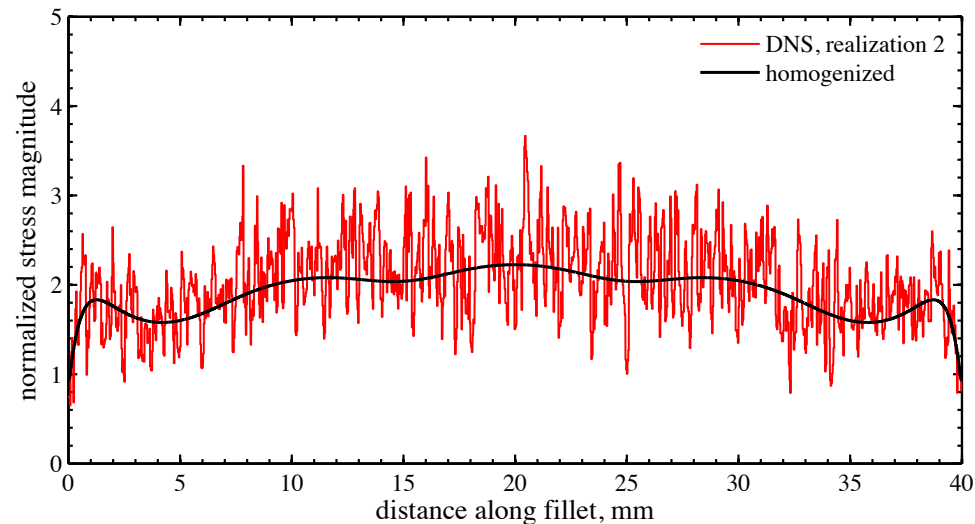
# Homogenized solution vs. DNS

## Stress magnitude along lower fillet

realization 1

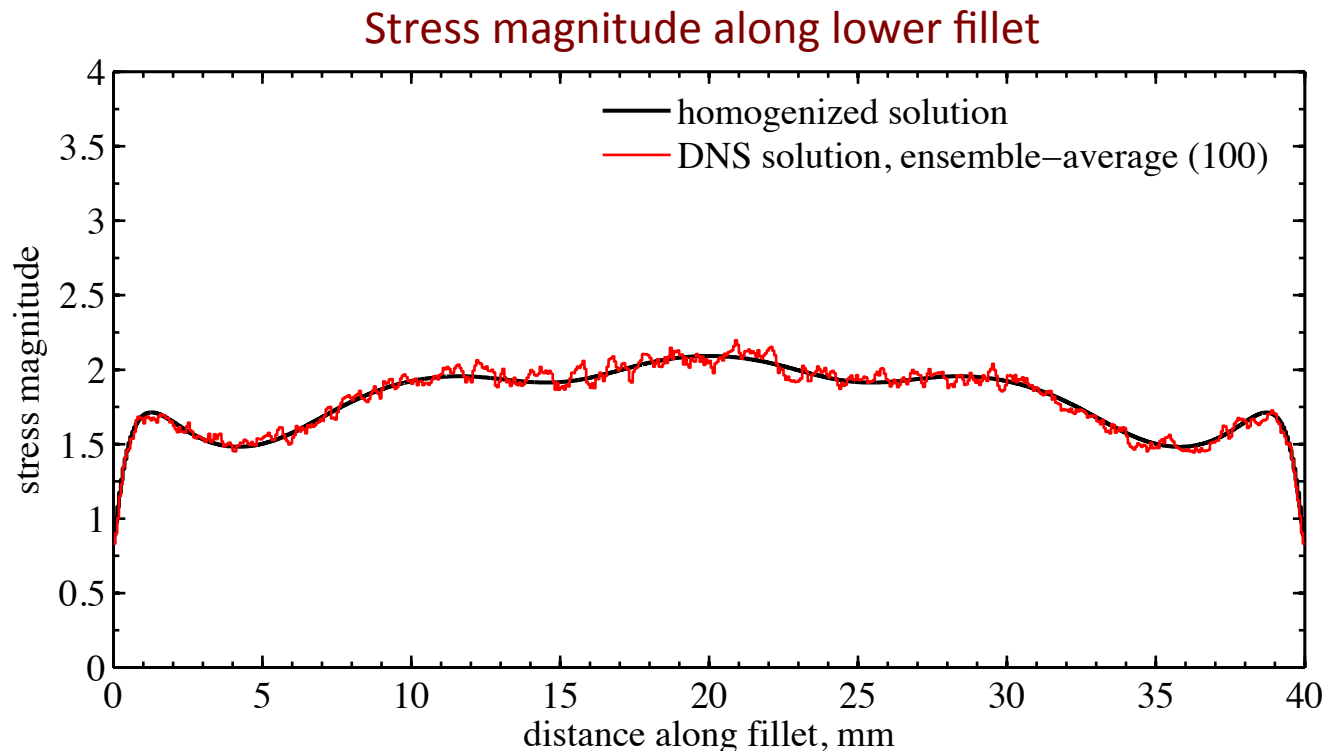


realization 2



# Homogenized solution vs. ensemble average

Beran and McCoy (1970) showed that the governing equation for the mean field is **nonlocal**.

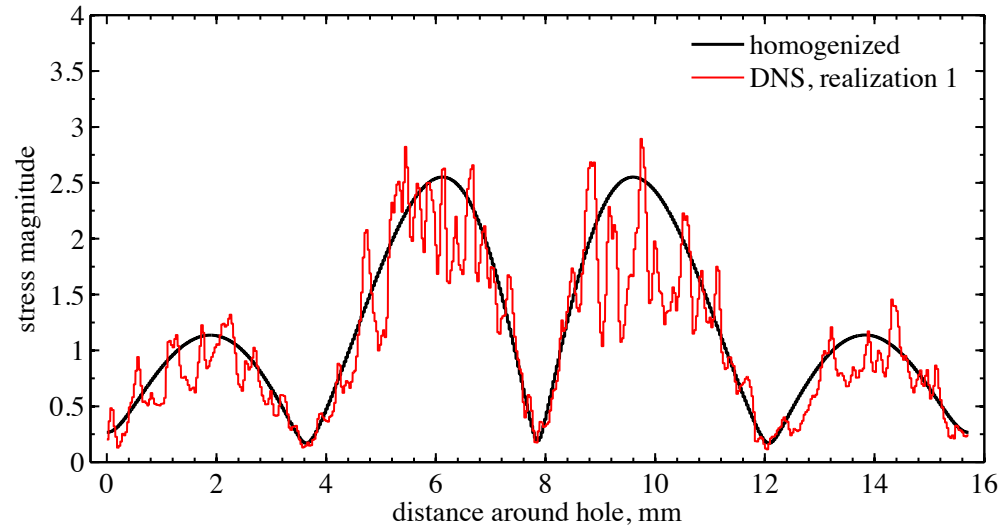


See no evidence for nonlocality here.

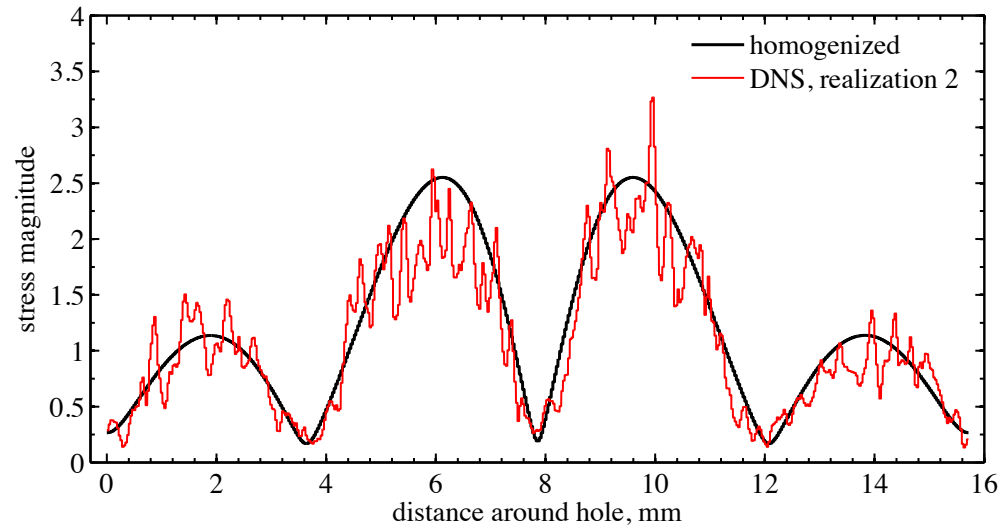
# Homogenized solution vs. DNS

## Stress magnitude around hole

realization 1

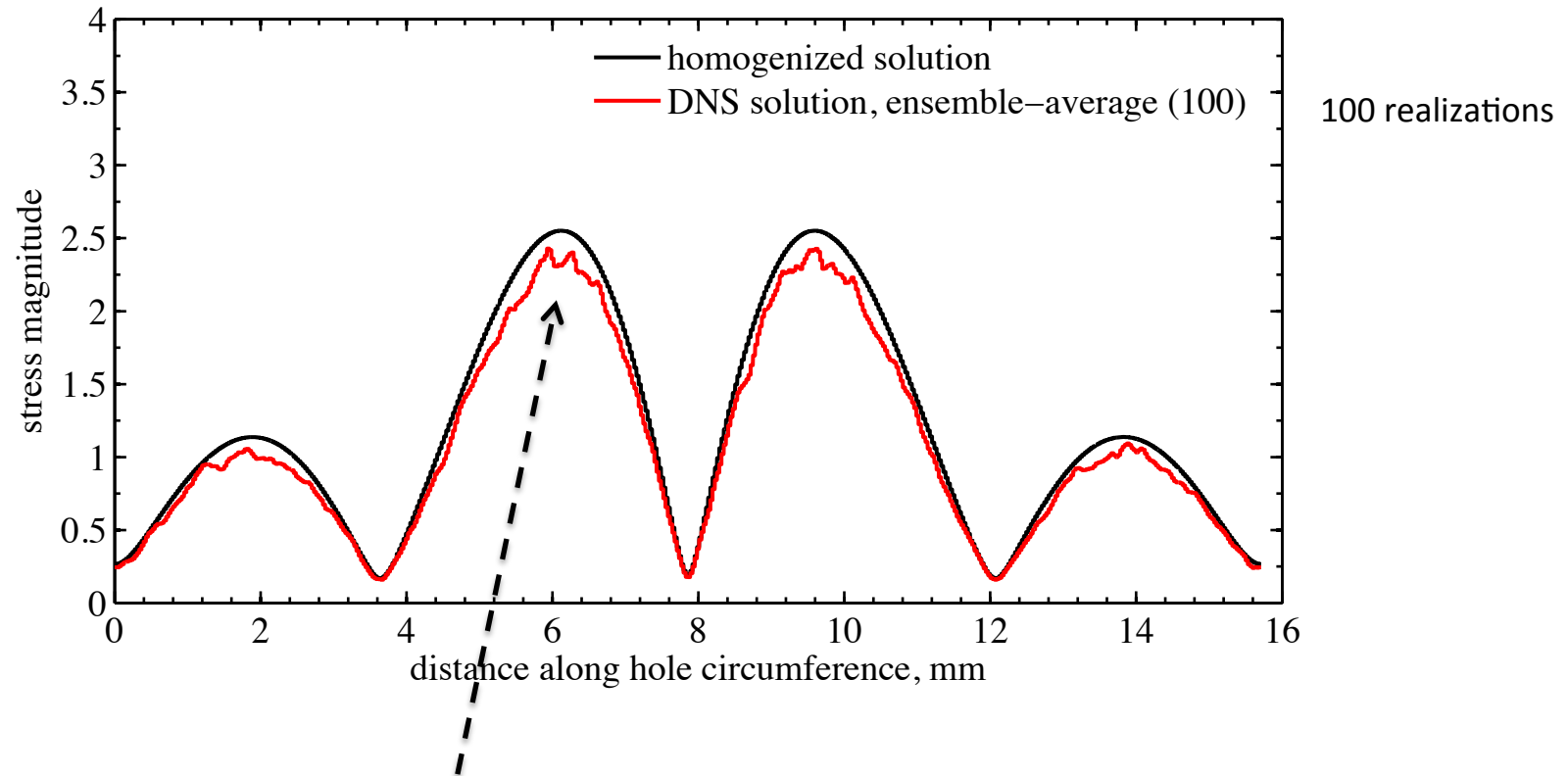


realization 2



# Homogenized solution vs. ensemble average

## Stress magnitude around hole



See some evidence for nonlocality here.

# 3D moving average using Gaussian filter

convolution

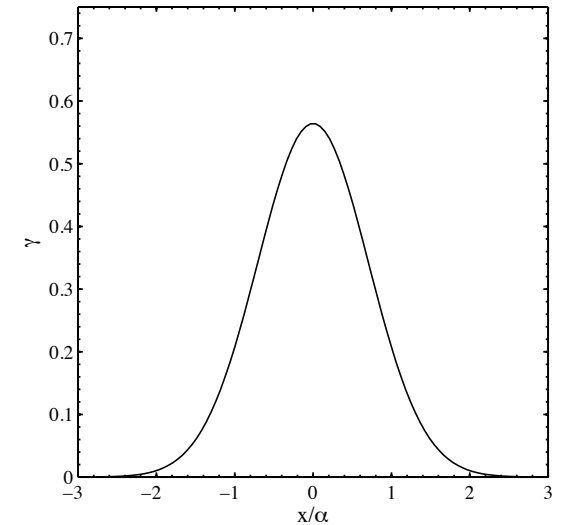
$$\hat{\sigma}_{ij}(\mathbf{x}) = \gamma_{\alpha}(\mathbf{x}) * \sigma_{ij}(\mathbf{x}) = \int_{\Omega_{\infty}} \gamma_{\alpha}(\mathbf{x} - \mathbf{y}) \sigma_{ij}(\mathbf{y}) d\mathbf{y}$$

$$\gamma_{\alpha}(\mathbf{x} - \mathbf{y}) = A e^{-\frac{||\mathbf{x} - \mathbf{y}||^2}{\alpha^2}}$$

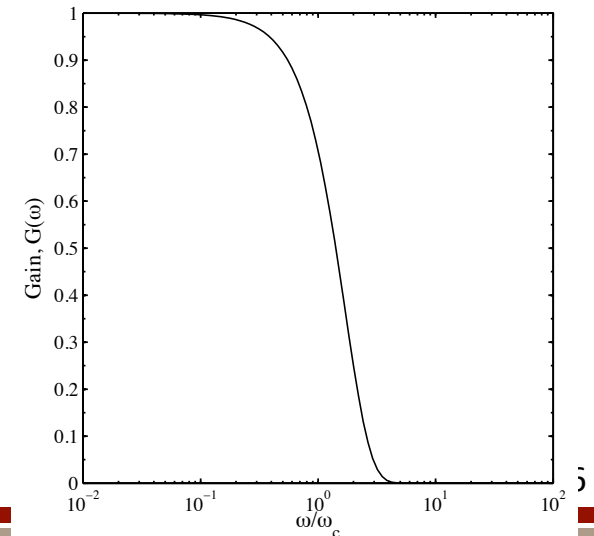
$A$  = normalization constant to reproduce constant functions

cutoff frequency  $\omega_c = \frac{\sqrt{2 \ln 2}}{\alpha} = \frac{1.1774}{\alpha}$

Gaussian Kernel



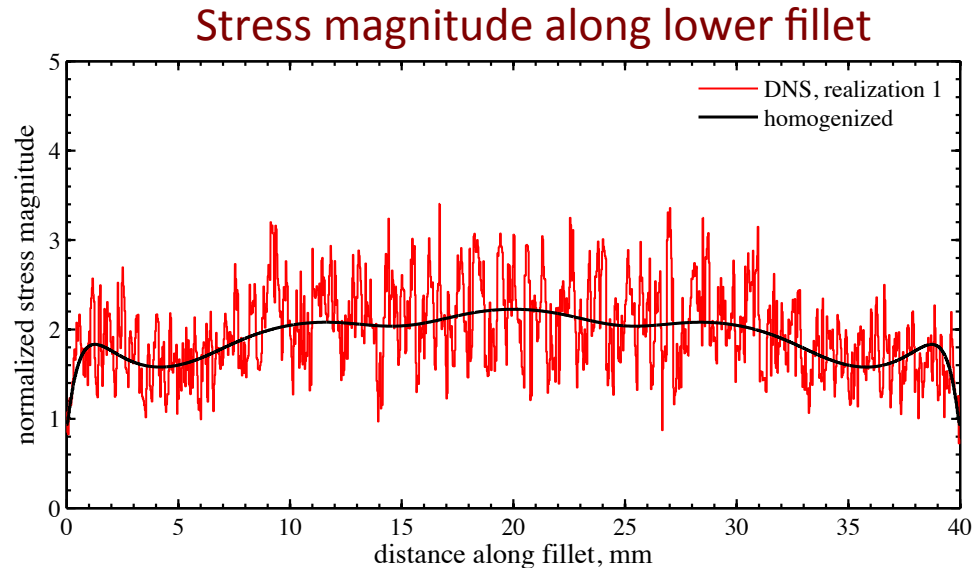
Gain vs. spatial frequency





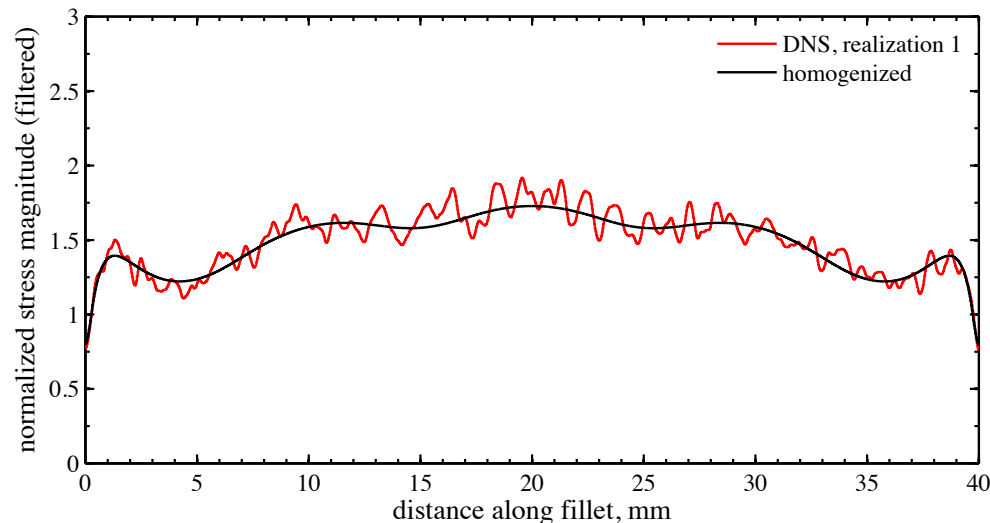
# 3D moving average using Gaussian filter

unfiltered



filtered

$\alpha = 0.125$  mm  
(moving average  
over  $\sim 2 \times 2 \times 2$  grains)

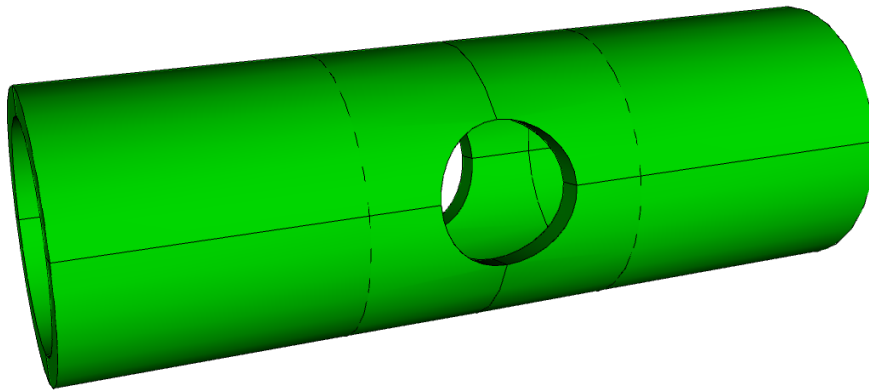


Homogenized  
solution is a  
surprisingly good  
approximation.

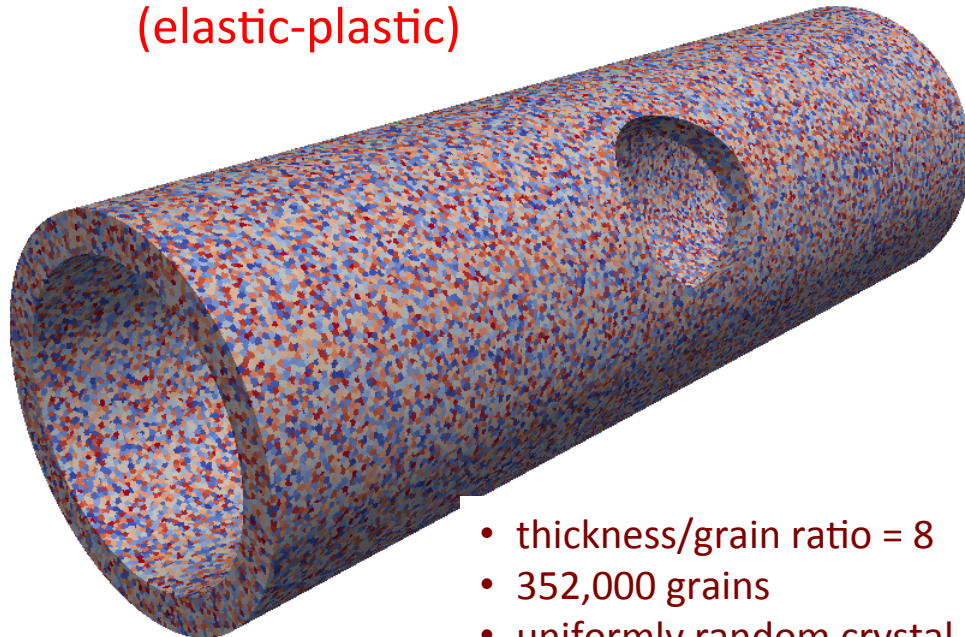
# Summary (elastic results)

- Found little evidence of higher-order effects for this material and these BVPs. This is possibly due to the small correlation length inherent in the microstructure.
- Fluctuations (10-20%) on the length scale of several grains are present as evidenced by spatially filtered DNS results.
- What about plastic regime?

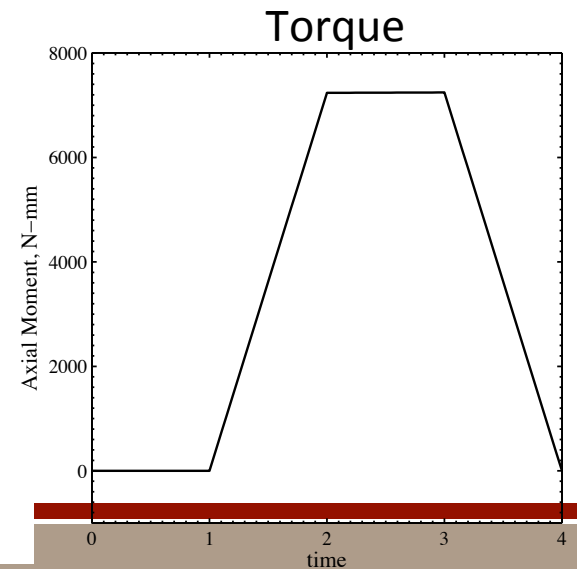
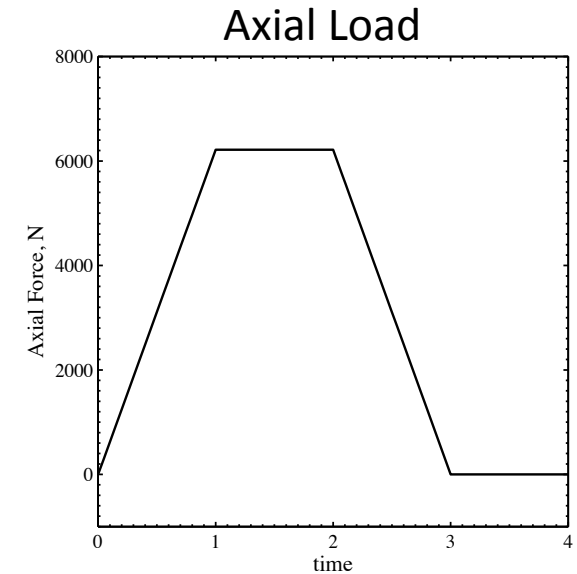
# Plastic example: stainless-steel tube under combined tension-torsion



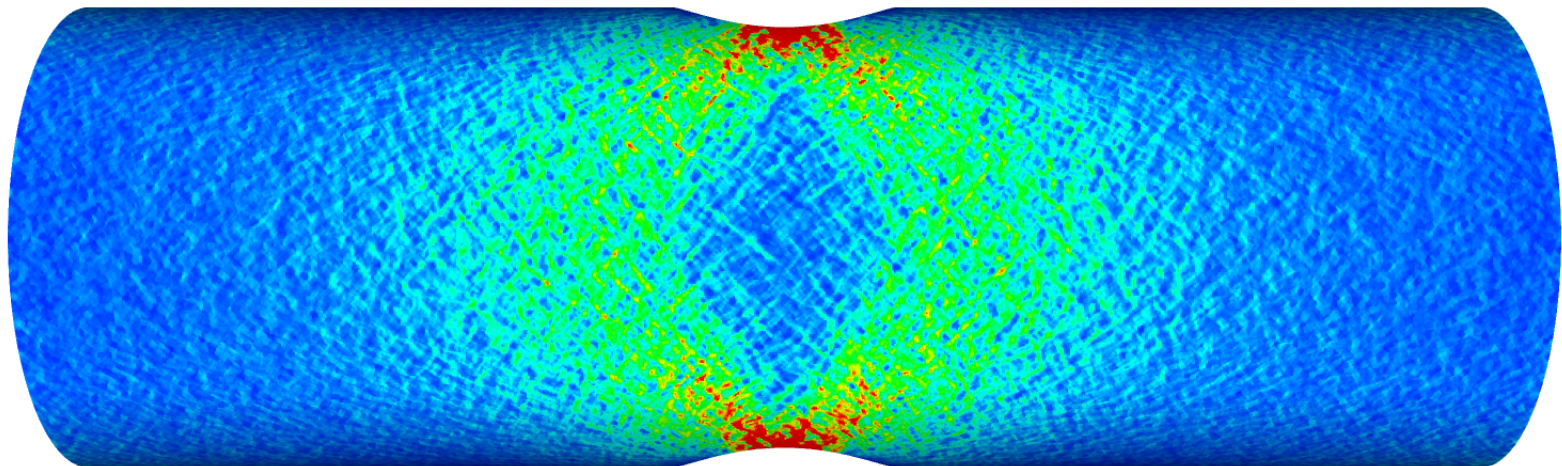
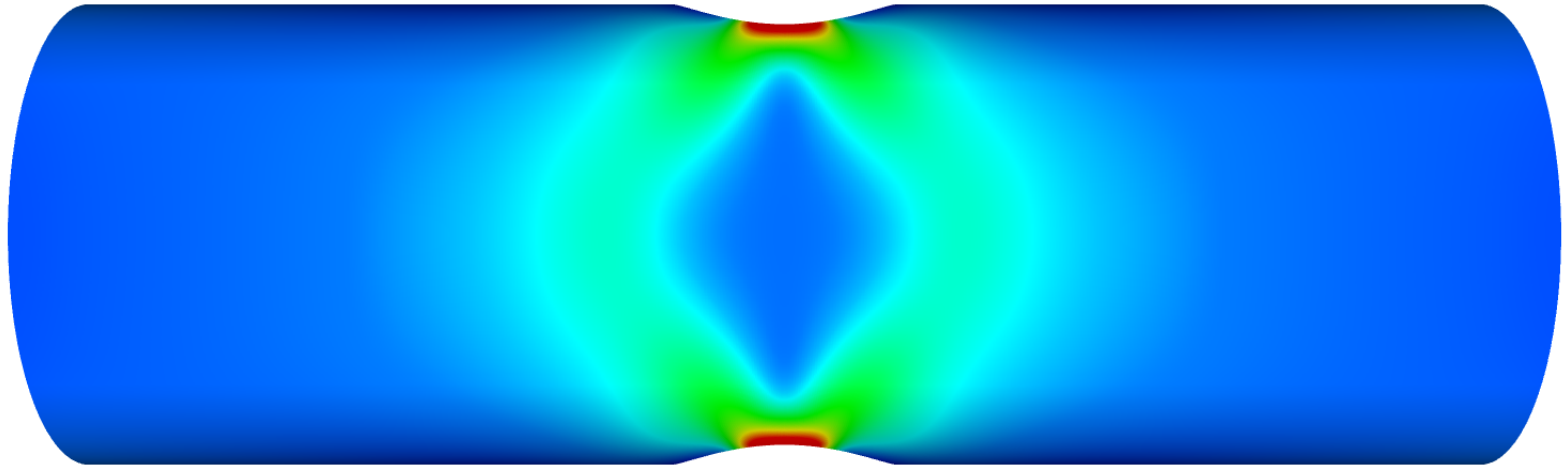
(elastic-plastic)



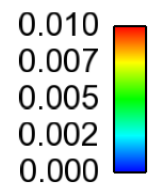
- thickness/grain ratio = 8
- 352,000 grains
- uniformly random crystal orientations



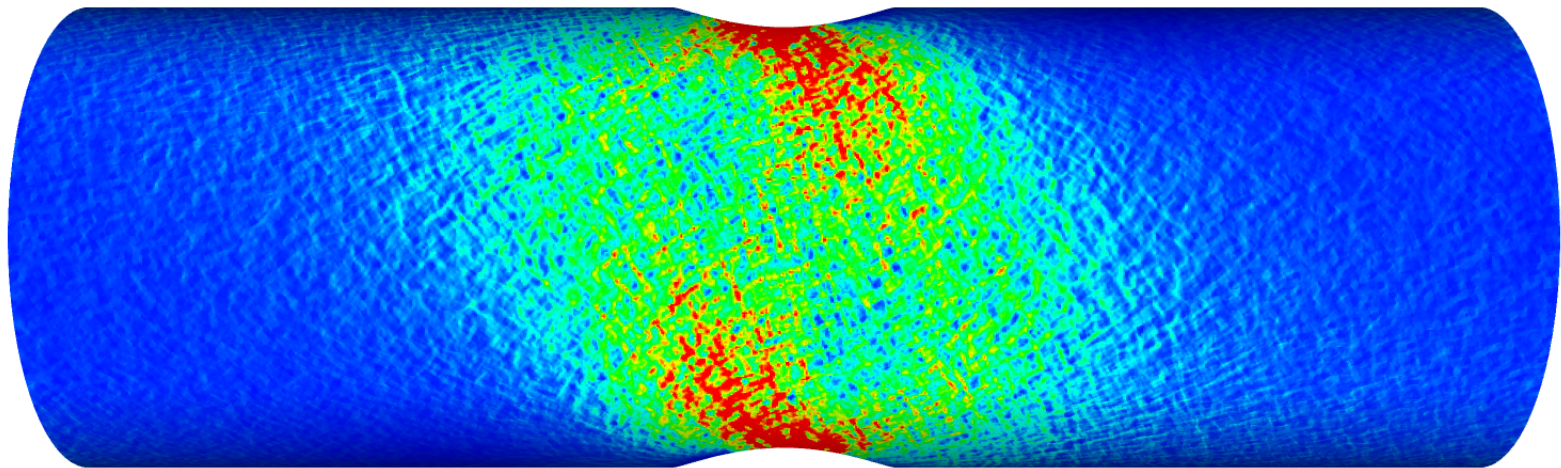
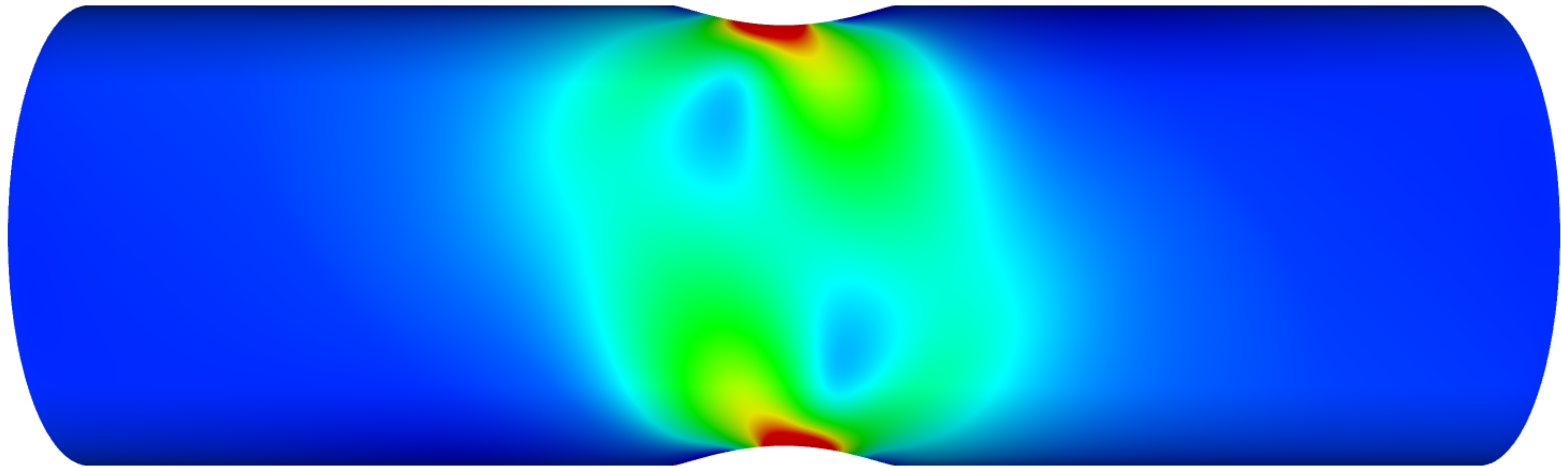
# Axial Load Only



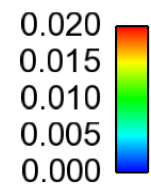
effective\_log\_strain



# Axial load + torsion

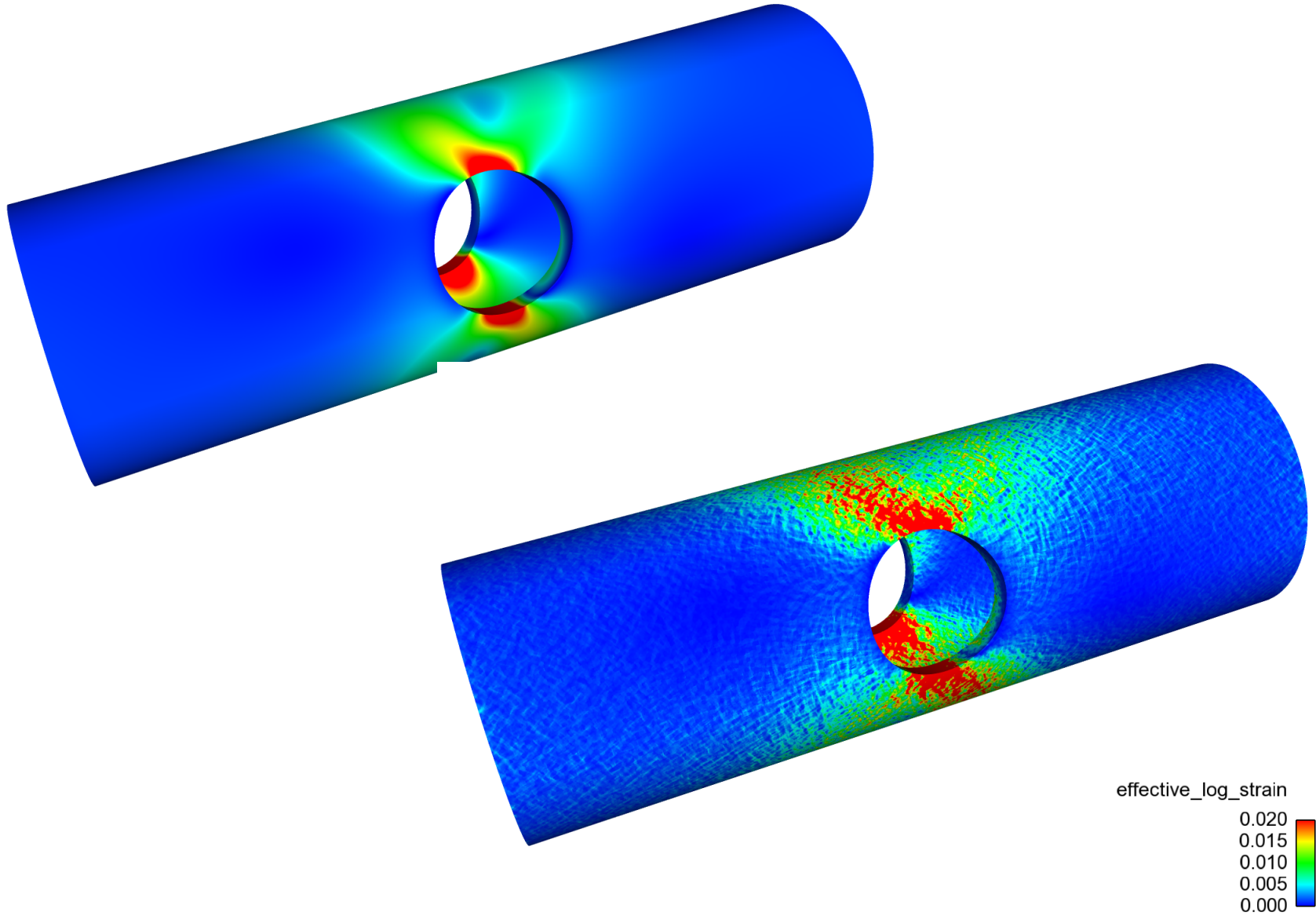


effective\_log\_strain



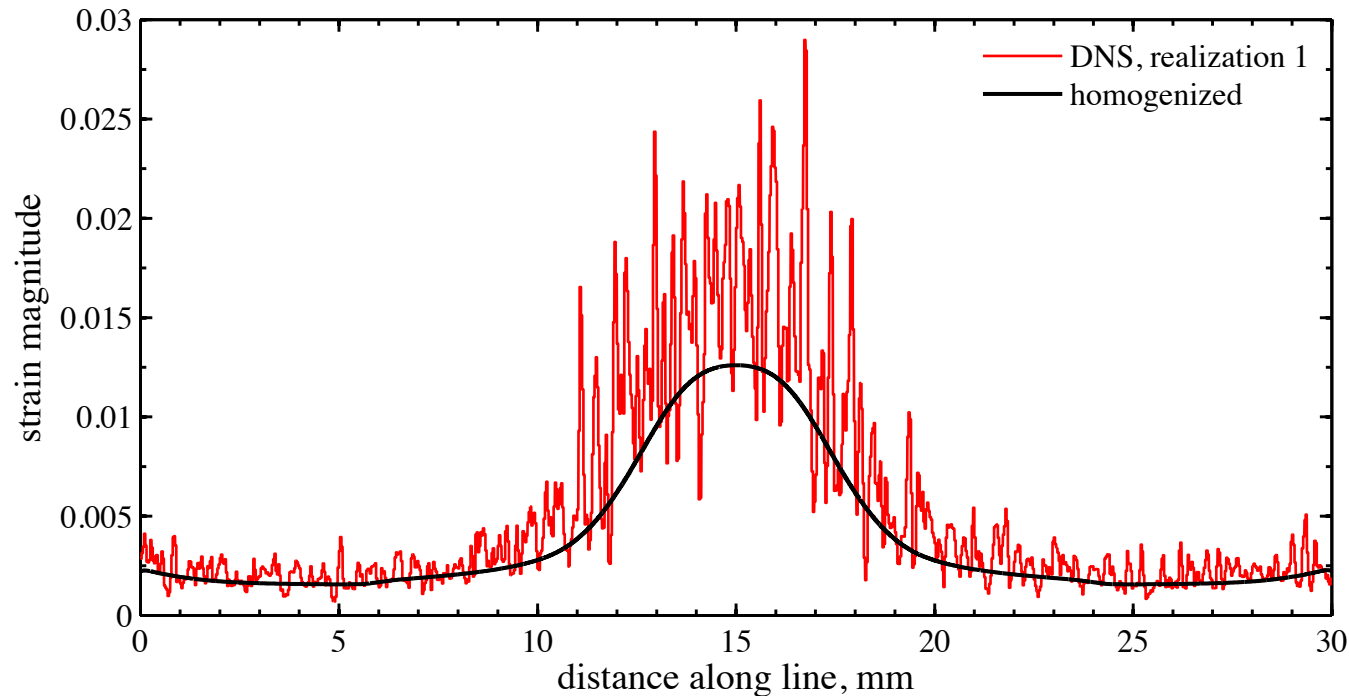


# Strain under combined tension-torsion



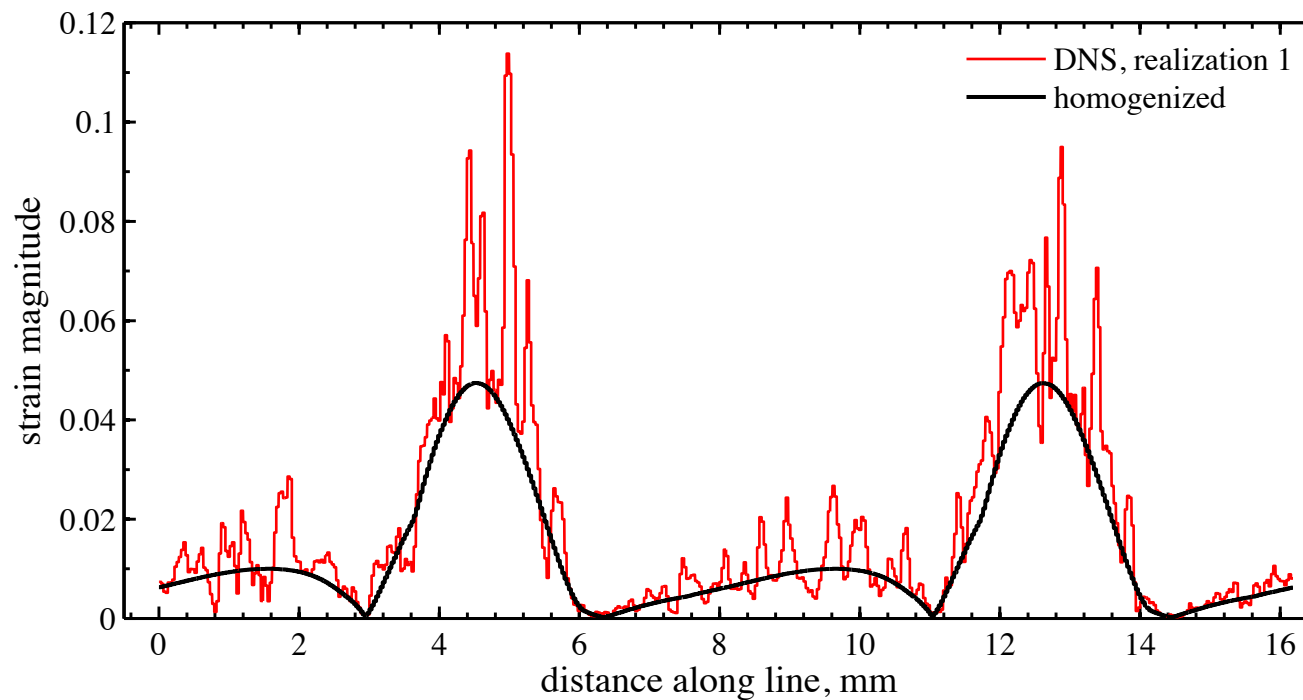
# Strain magnitude along length of tube

midsection between holes, combined tension-torsion



# Strain magnitude around hole

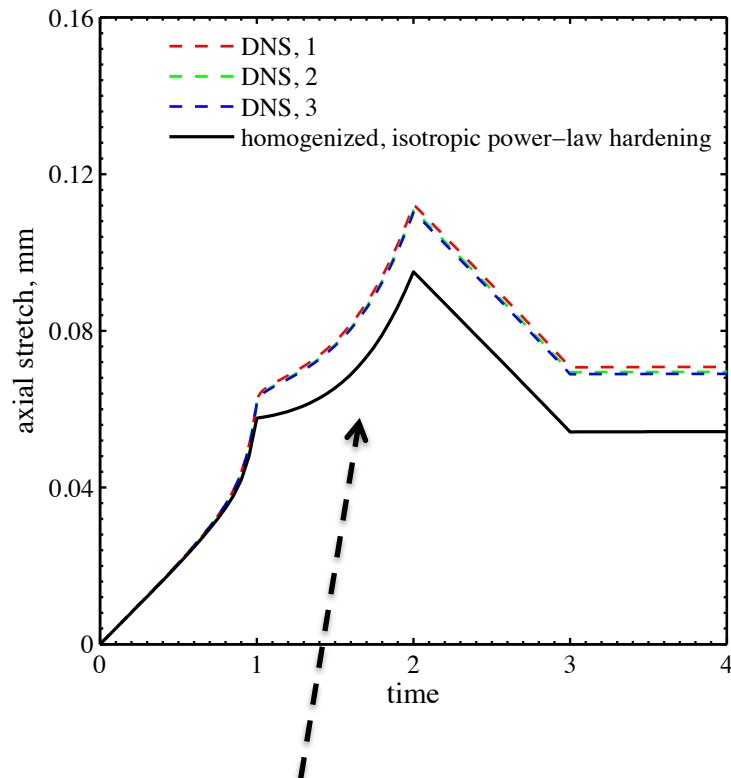
inside circumference, combined tension-torsion



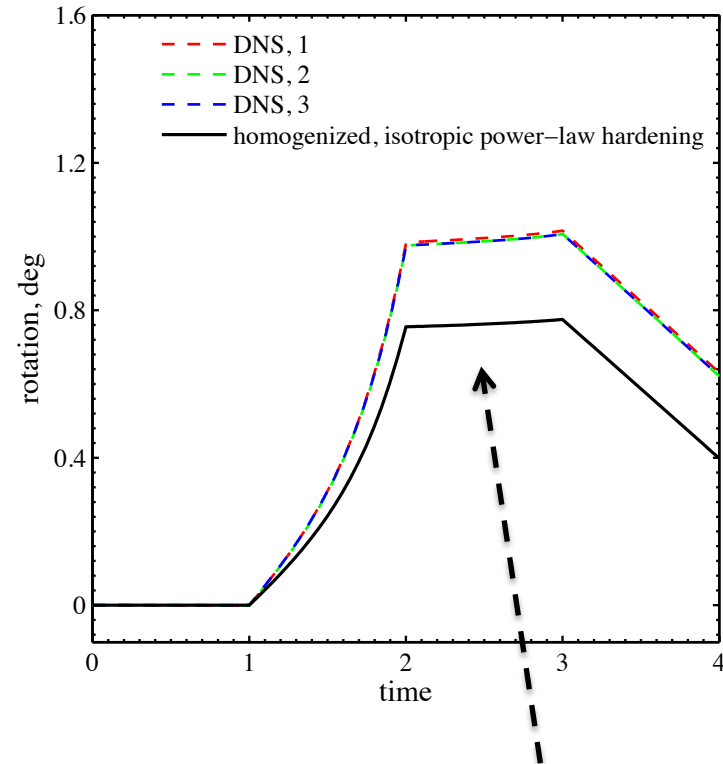


# Global stretch and rotation of tube

axial stretch



rotation



Homogenized solution good in tension-only region but less accurate in combined tension-torsion.

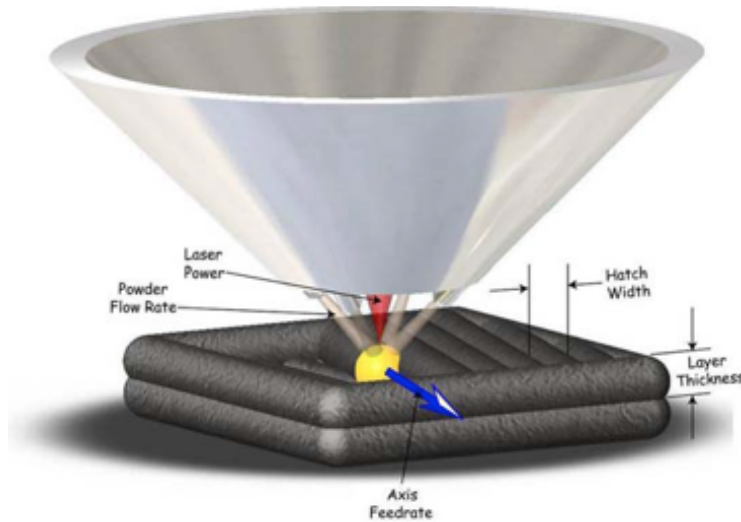
# Summary (plastic results)

- See appreciable difference between a basic J2 plasticity model and DNS results.
- Need full FE<sup>2</sup> for true homogenization in the plastic regime.
- What about more complex microstructures, e.g. from additive?

Bishop, J., Emery, J., Field, R., Weinberger, C., Littlewood, D. 2015, "Direct numerical simulations in solid mechanics for understanding the macroscale effects of microscale material variability," *CMAME*, 287, pp. 262-289.

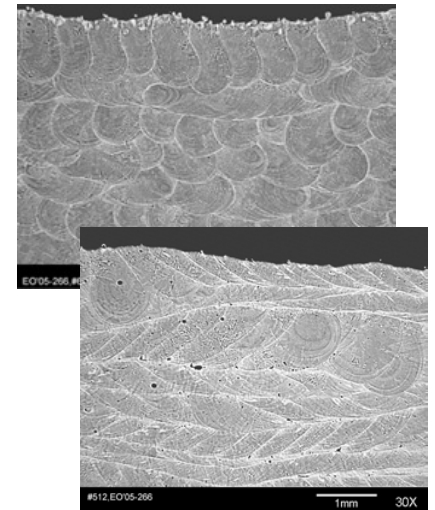
# Application to additive manufacturing

# Laser Engineered Net Shape (LENS)



Schematic of LENS™ laser-based deposition process

- LENS “hatch” structure results in a complex mesoscale structure.
- Classical assumption of scale-separation may no longer be applicable.



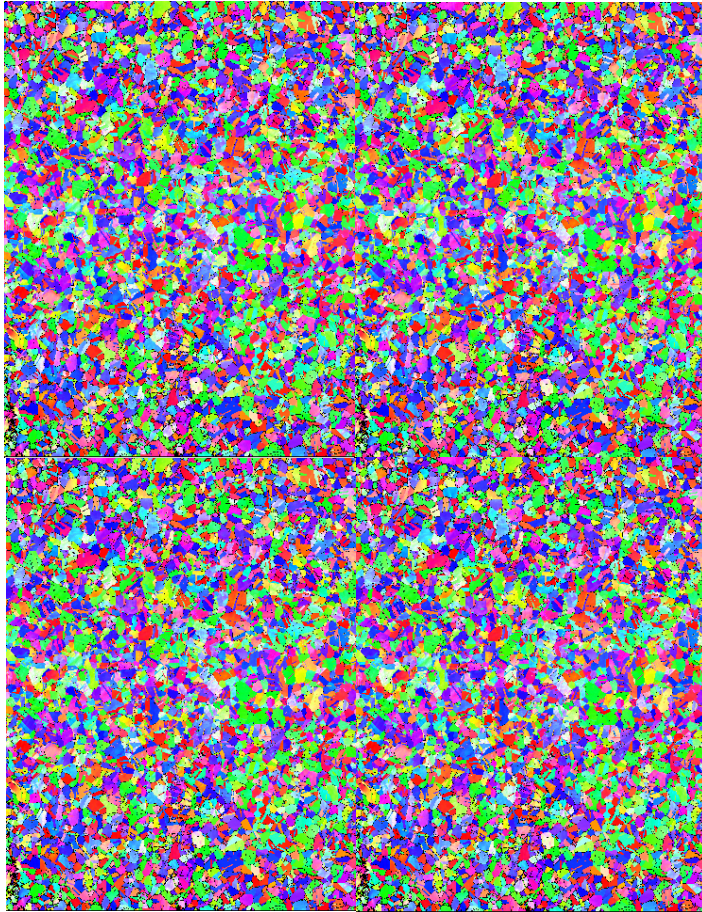
LENS mesostructure

# Microstructure: wrought vs. LENS

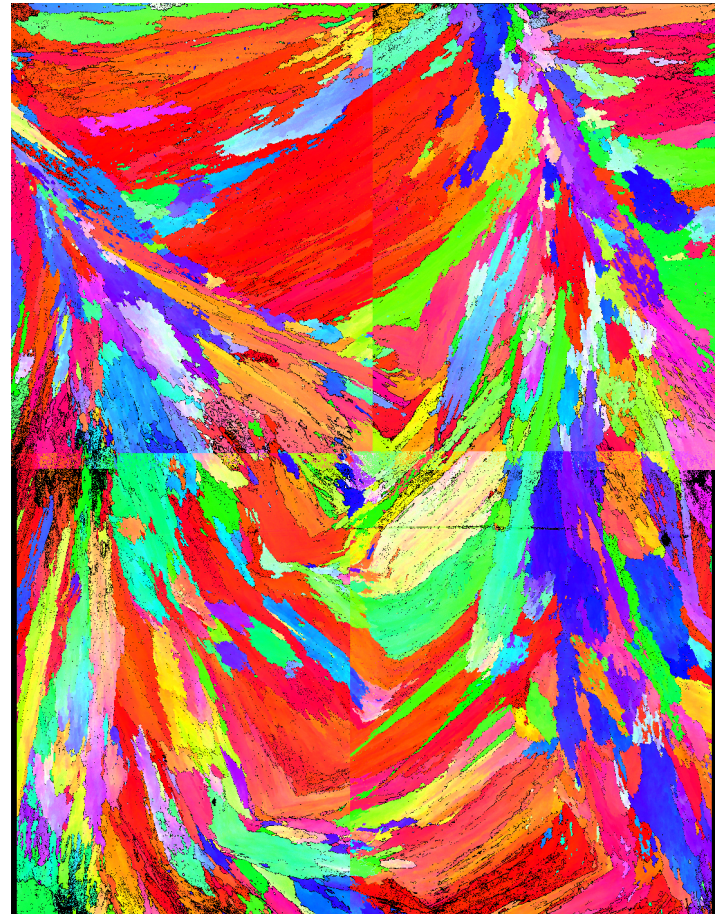
Wrought, SS 304L

(Images shown  
at same scale.)

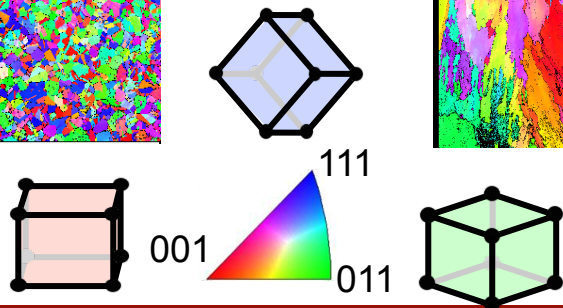
LENS, SS 304L, 3.8 kW



1.0 mm

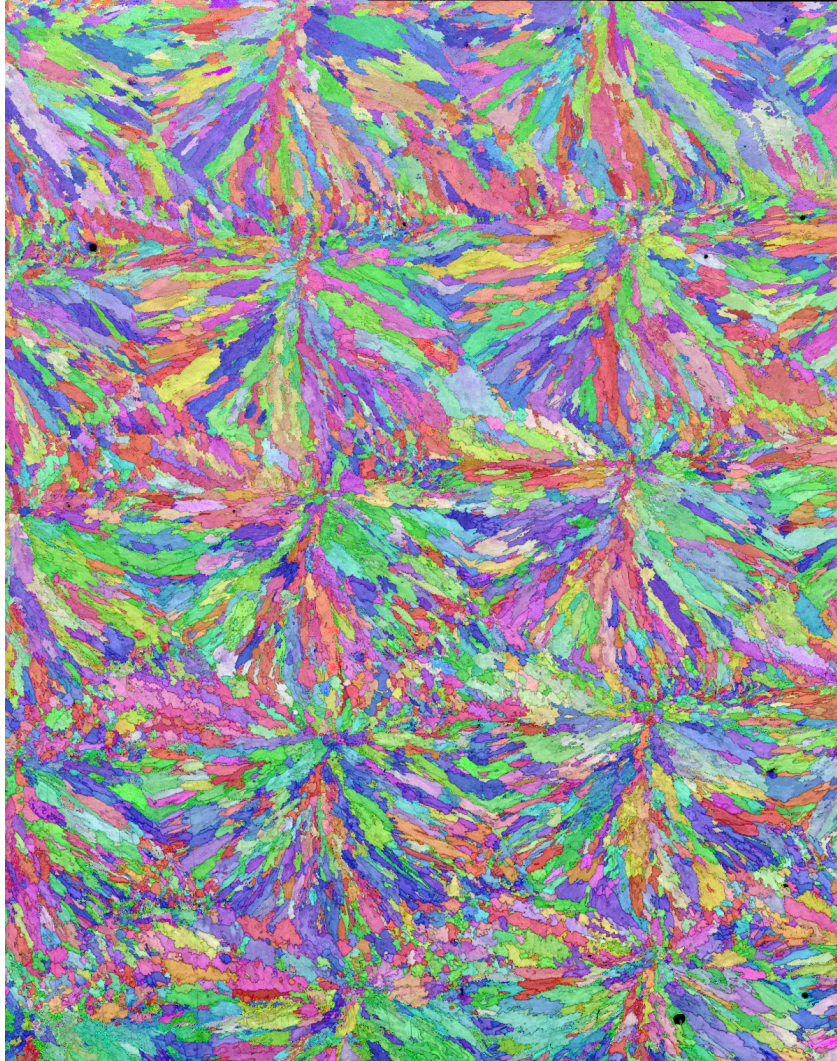


1.0 mm



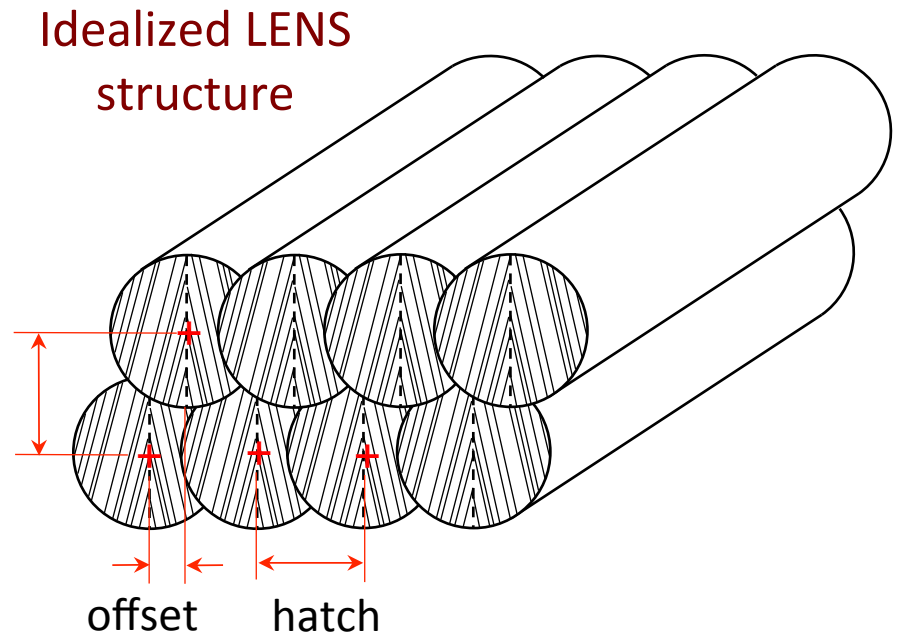
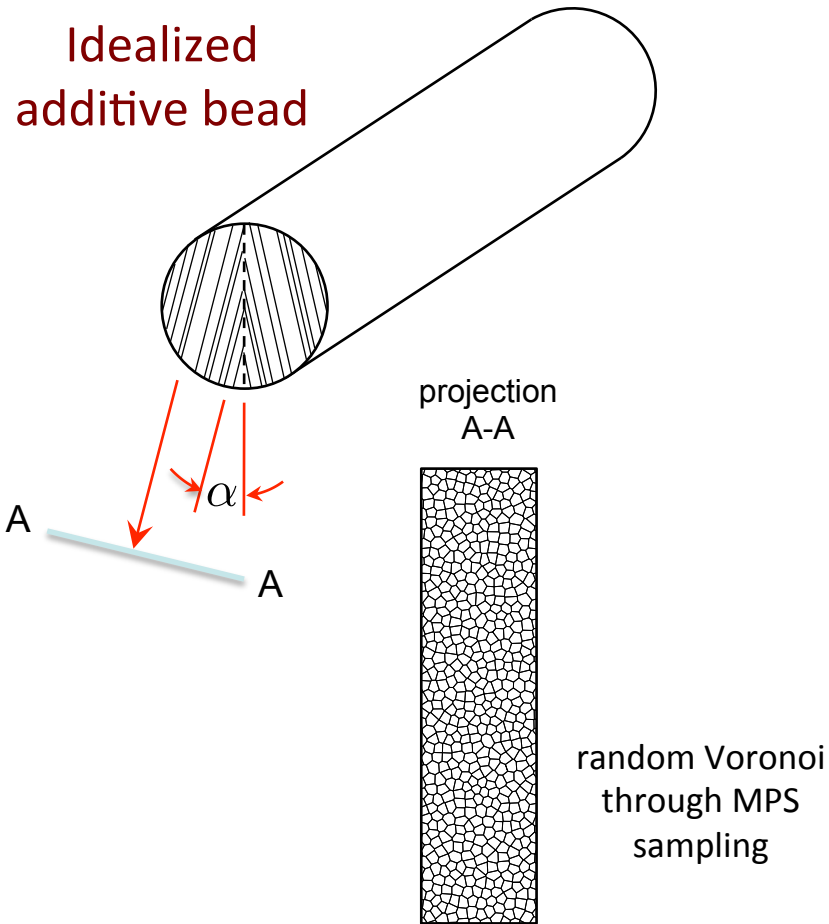


# LENS microstructure



8 mm x 10 mm

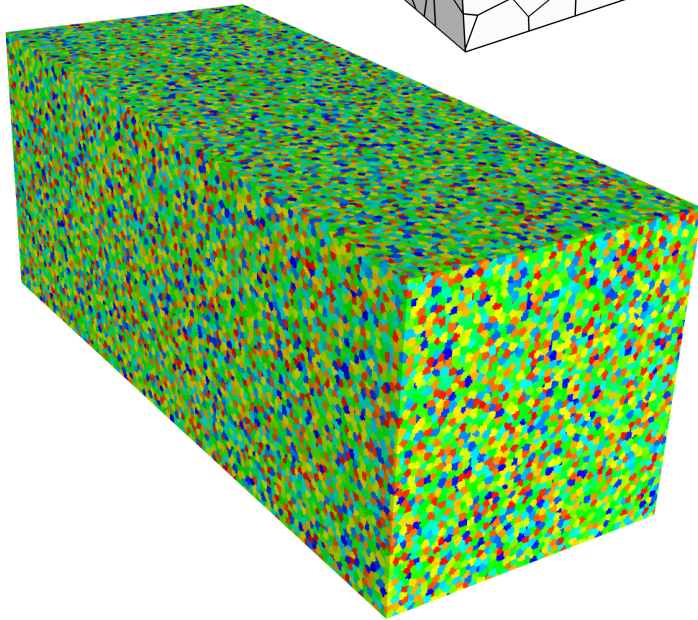
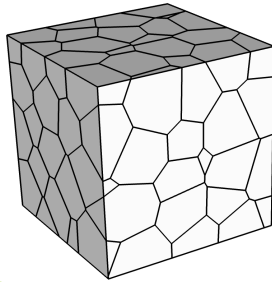
# Idealized LENS microstructures



texture?

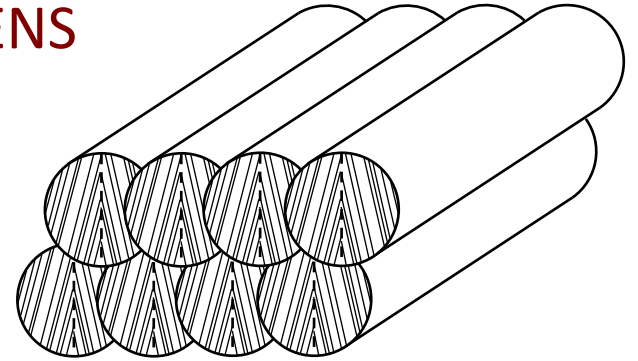
# DNS modeling

equiaxed



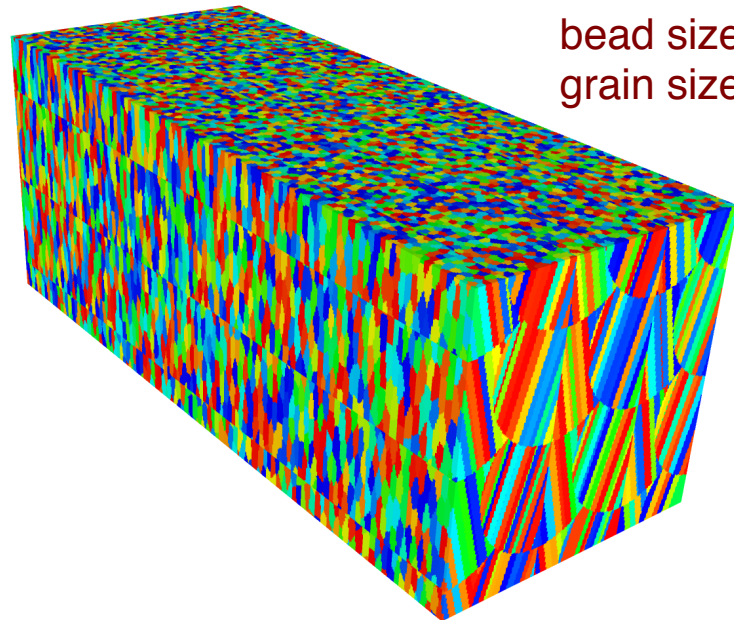
grain size = 40 microns

additive, LENS



bead size = 1 mm

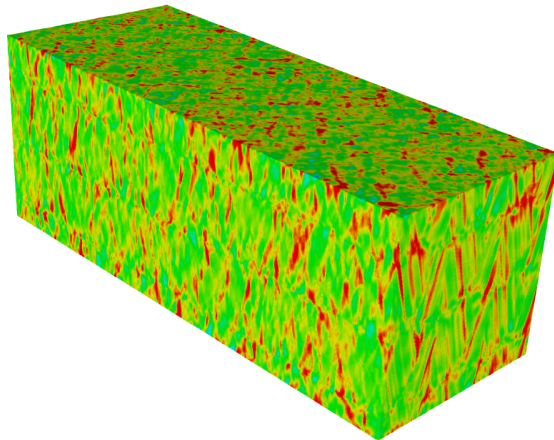
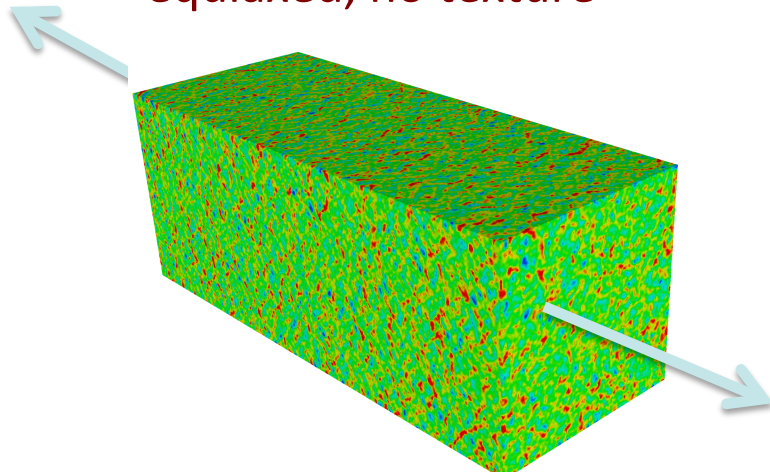
grain size = 40 microns



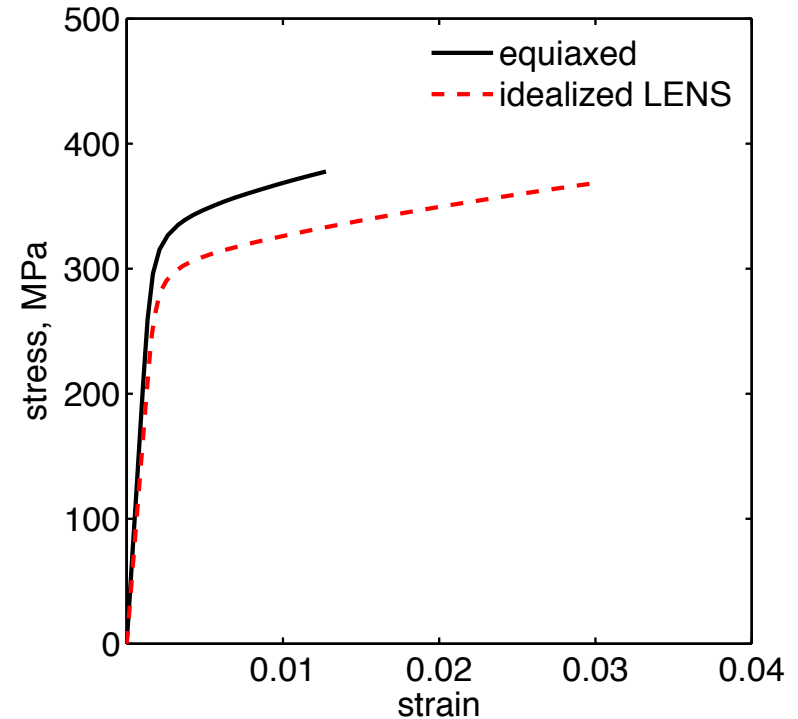


# Engineering stress-strain

equiaxed, no texture

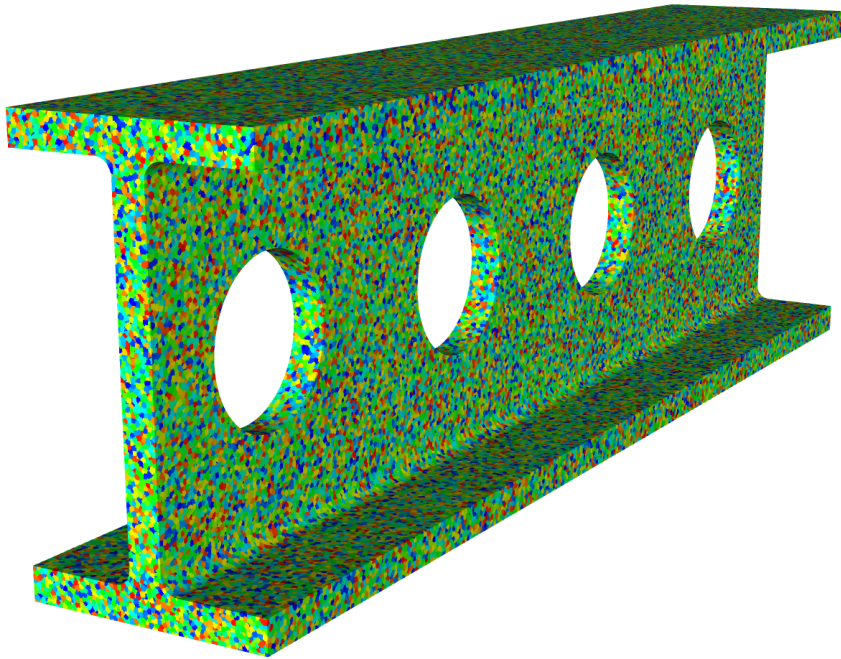


additive, LENS

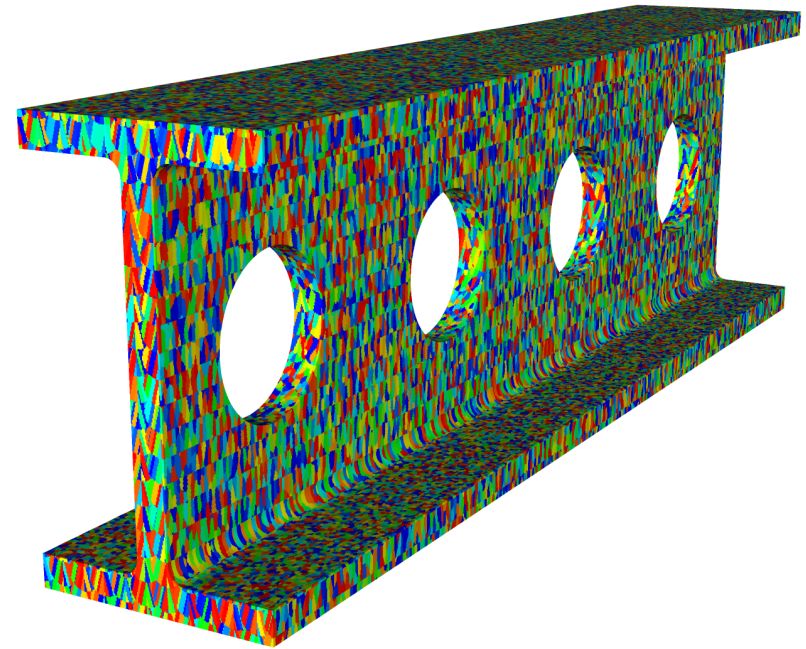


# Idealized microstructures

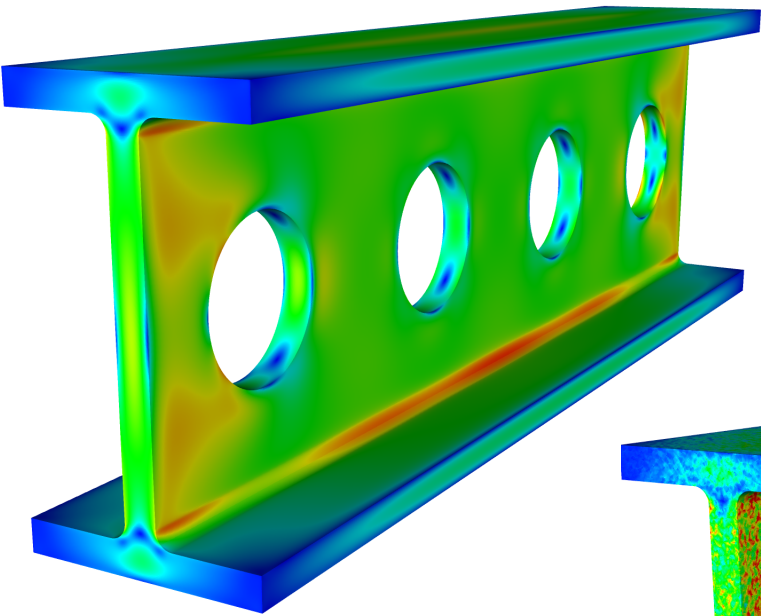
equiaxed



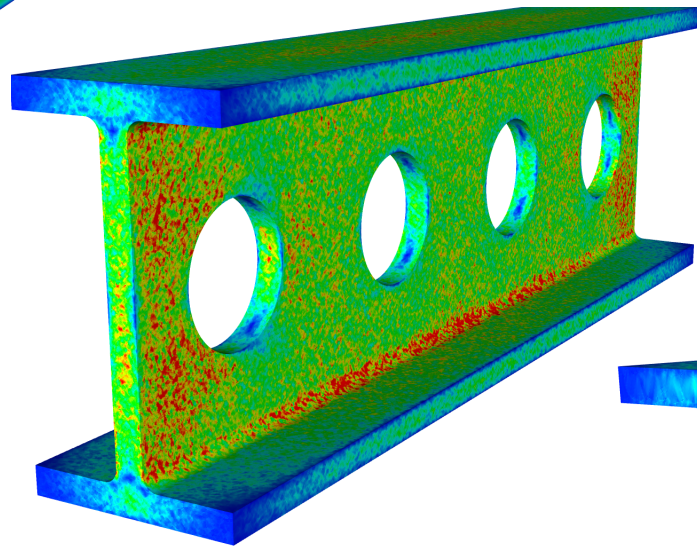
LENS



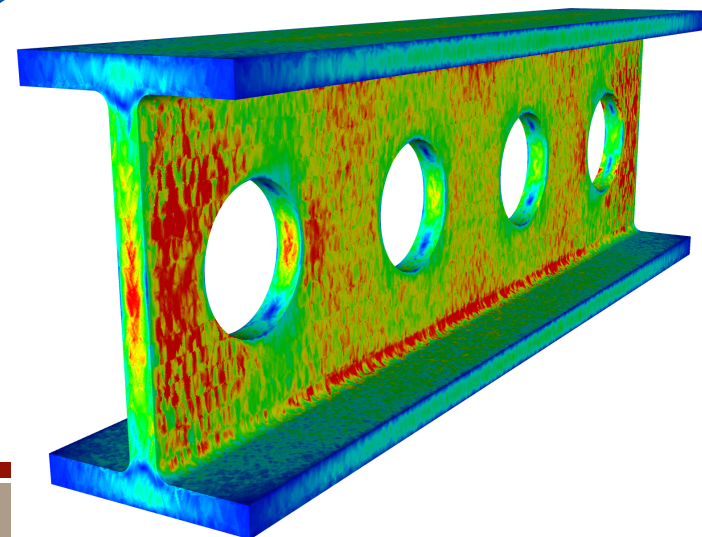
# Macroscopic stress field



homogeneous, isotropic



equiaxed, no texture, isotropic



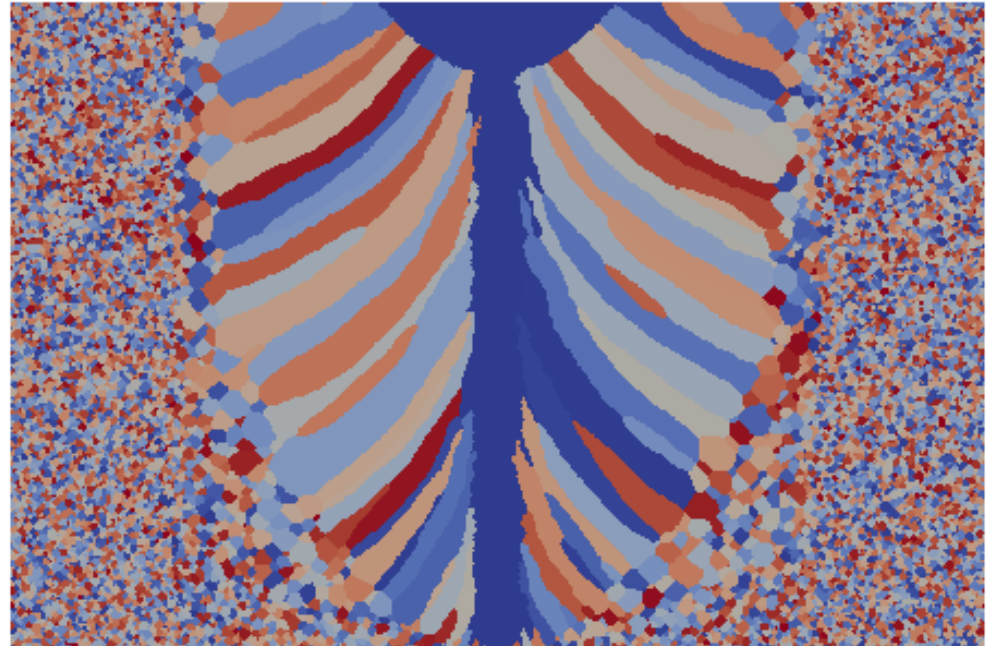
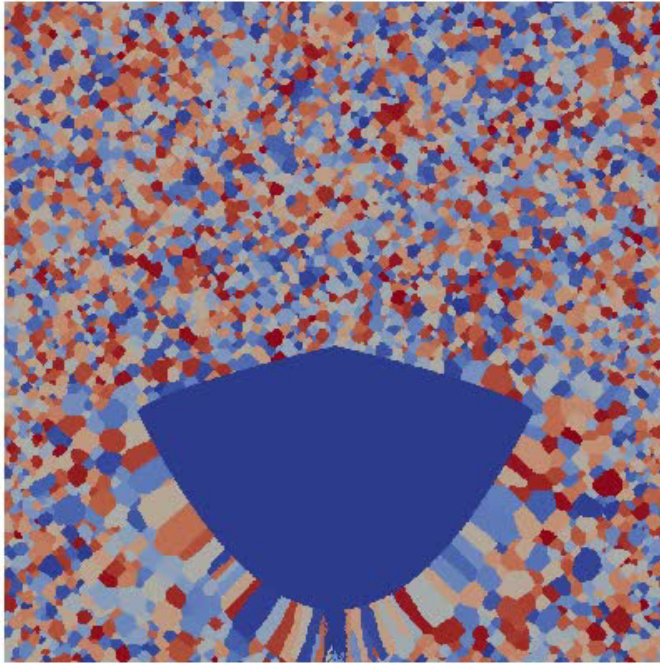
idealized LENS



# Process modeling

(Veena Tikare, SNL)

Future tie-in with process modeling:  
grain growth simulation



Kinetic Monte Carlo

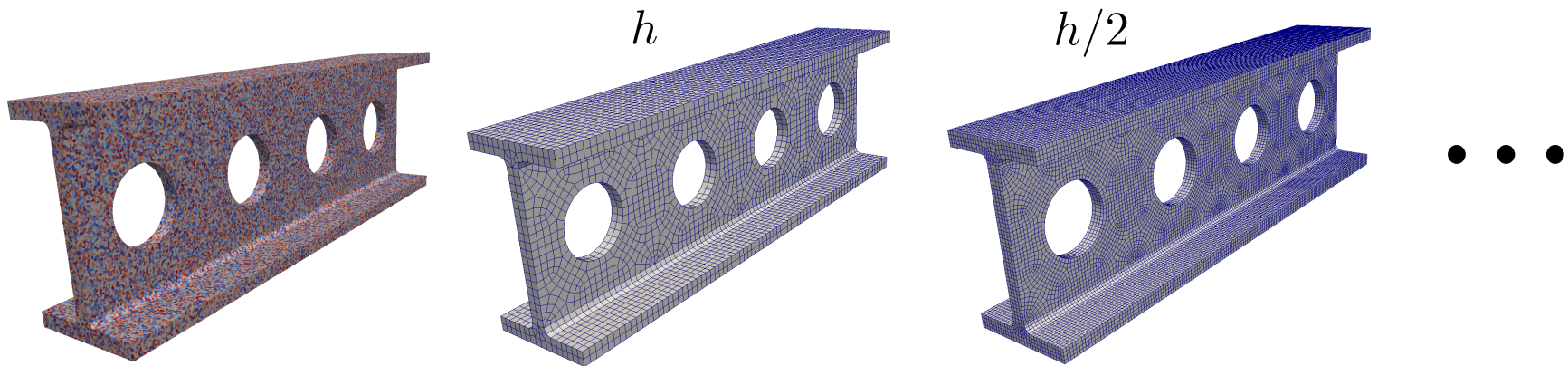
# Multiscale modeling using 'geometric multigrid' concepts

Miehe & Bayreuther, 2007

Fish & Belsky, 1995

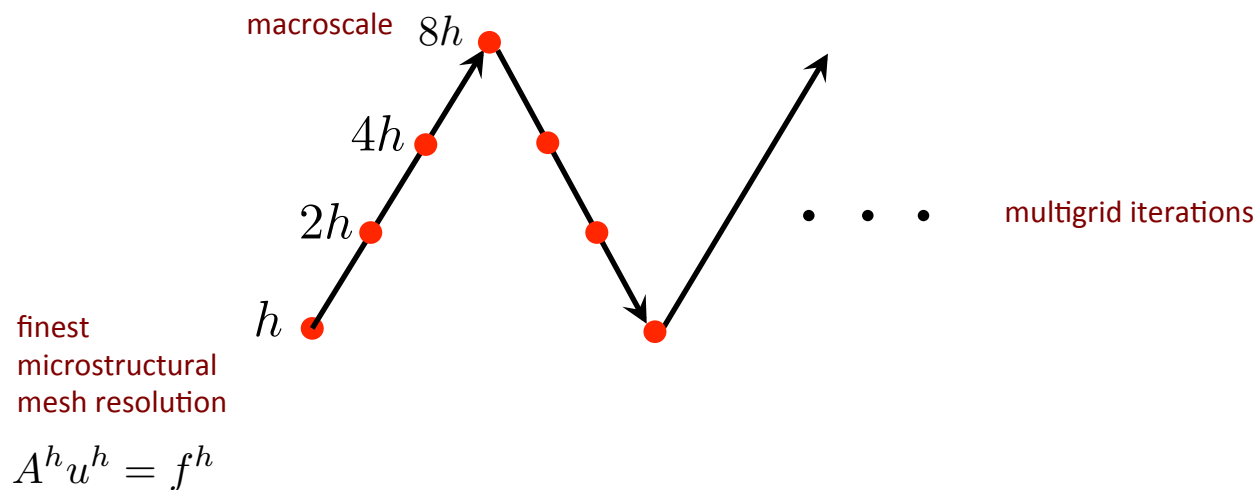
# Multiscale multigrid solvers

- Key postulate: we have a fine scale representation/model that is predictive (e.g. microstructural model, crystal-plasticity model)
- Key idea: use geometric multigrid concepts to create a multiscale method (solver) that optimally obtains the “solution” at all scales
- No assumption of scale separation
- Applicable to linear or nonlinear problems
- Use goal-oriented error estimation to optimally create multigrid hierarchy
- Each grid represents a filtering of the fine-scale physics



We have a natural hierarchy of grids in our DNS approach.

# Multigrid iterative V-cycle



$$u^h = I_{2h}^h u^{2h}$$

submodeling based prolongation

$$I_h^{2h} = (I_{2h}^h)^T$$

Galerkin property

$$f^{2h} = I_h^{2h} f^h$$

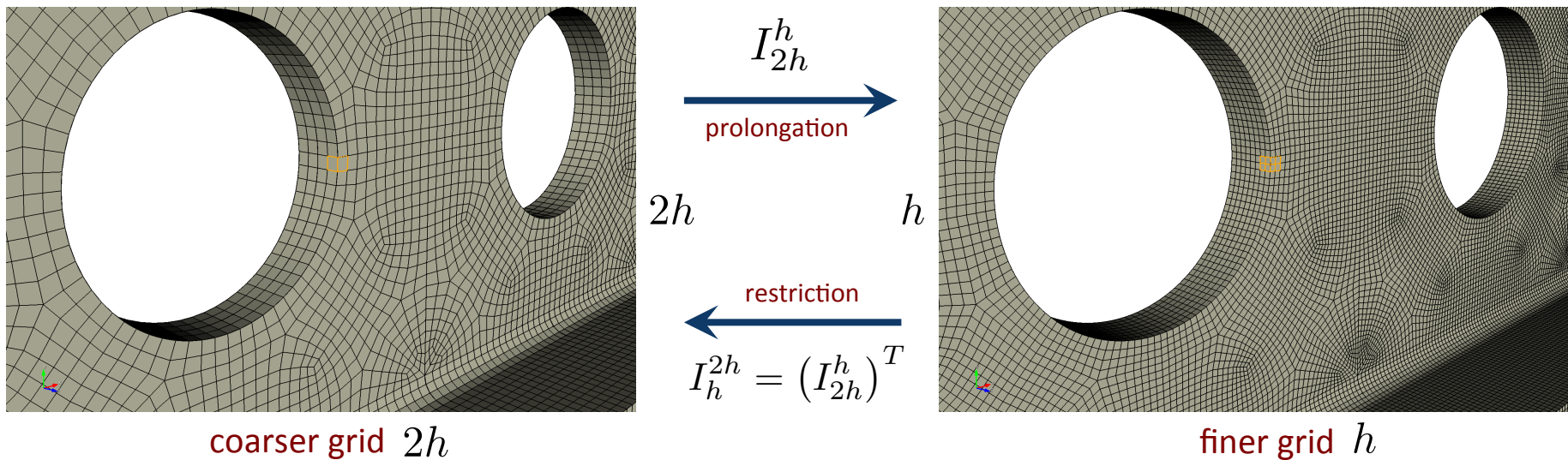
$$u^{2h} = I_h^{2h} u^h$$

$$A^{2h} = I_h^{2h} A^h I_{2h}^h$$

coarser grid operator

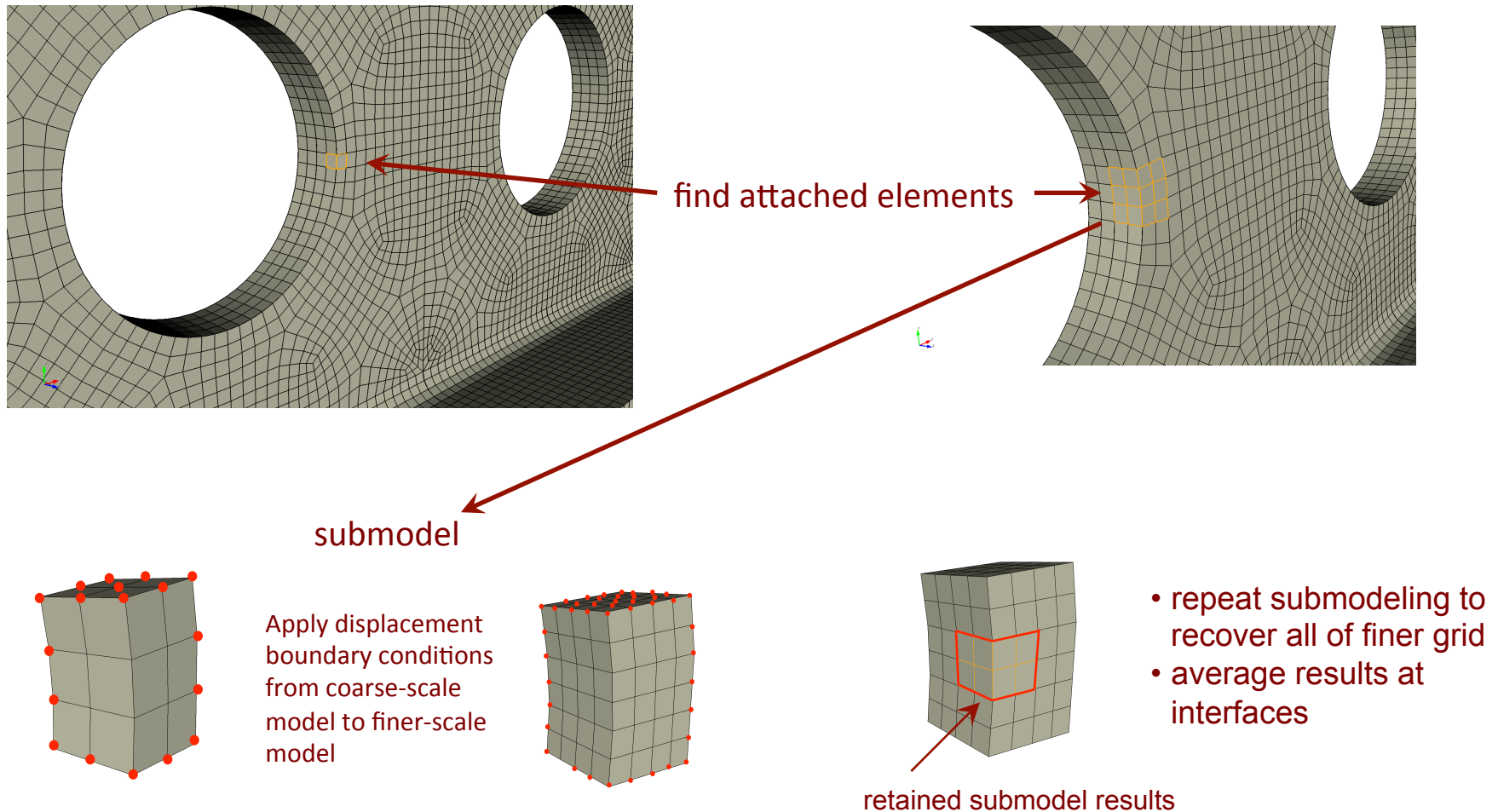
# Intergrid transfer operators? $I_{2h}^h$

- Intergrid transfer operators (prolongation and restriction) are key elements of geometric multigrid methods.
- Due to material heterogeneity, can NOT use standard prolongation and restriction operators.
- Use 'submodeling' techniques to define prolongation operator (coarse to fine).
- Restriction operator (fine to coarse) is given by variational optimality condition.





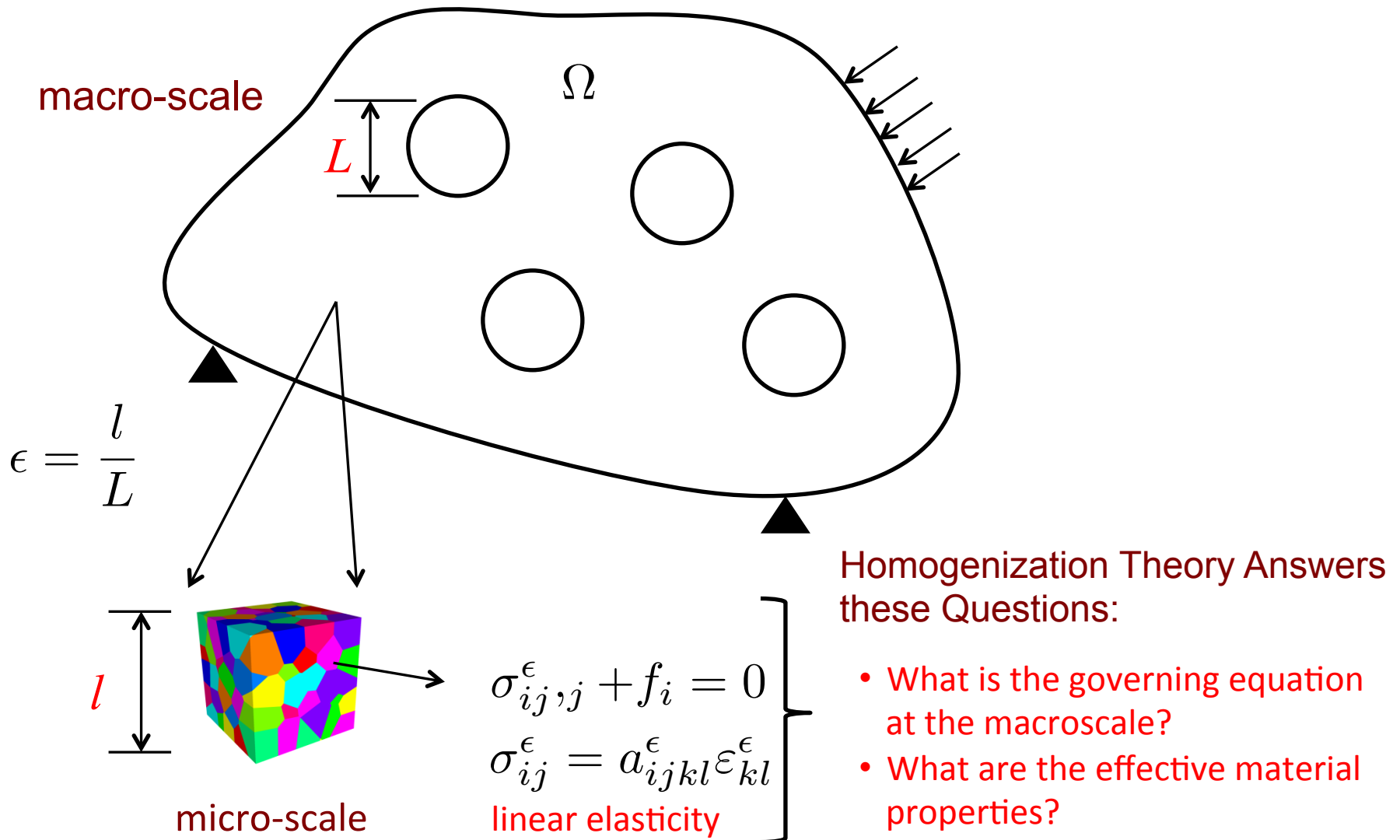
# Construct $I_{2h}^h$ using overlapping submodels



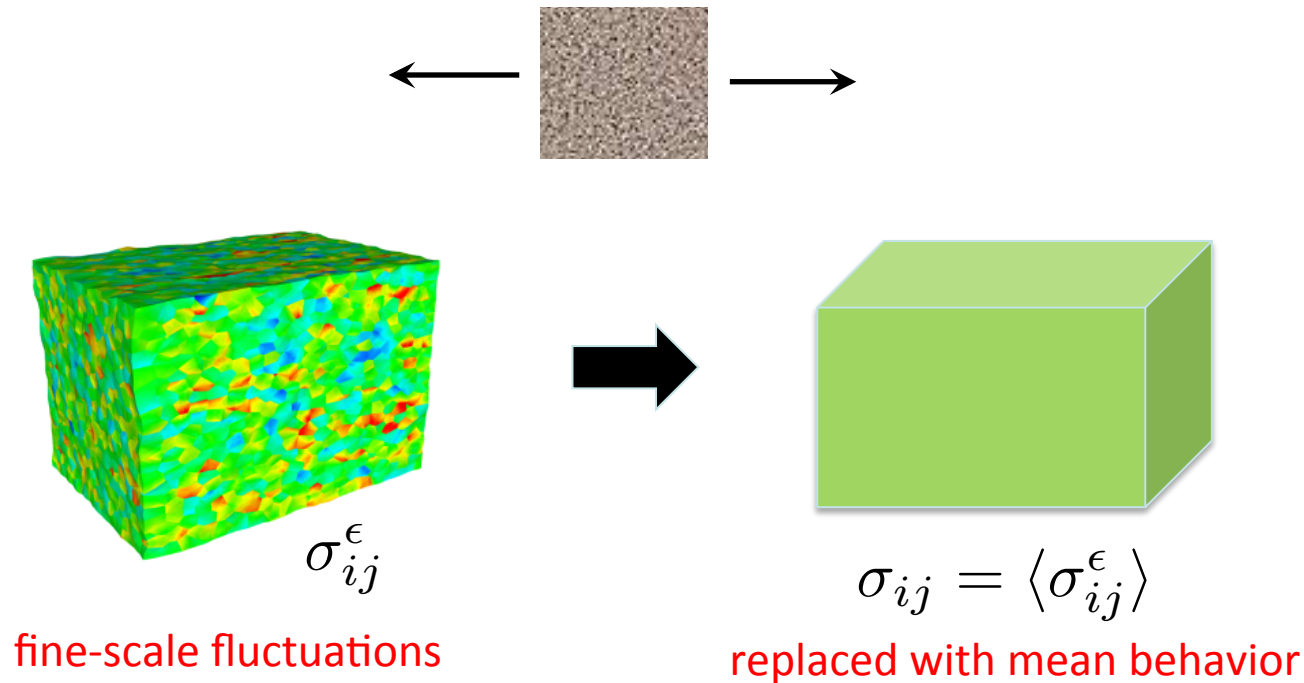
Thank you!

# Extra

# What about the Governing PDE?



# Homogenization



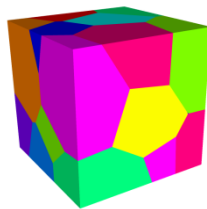
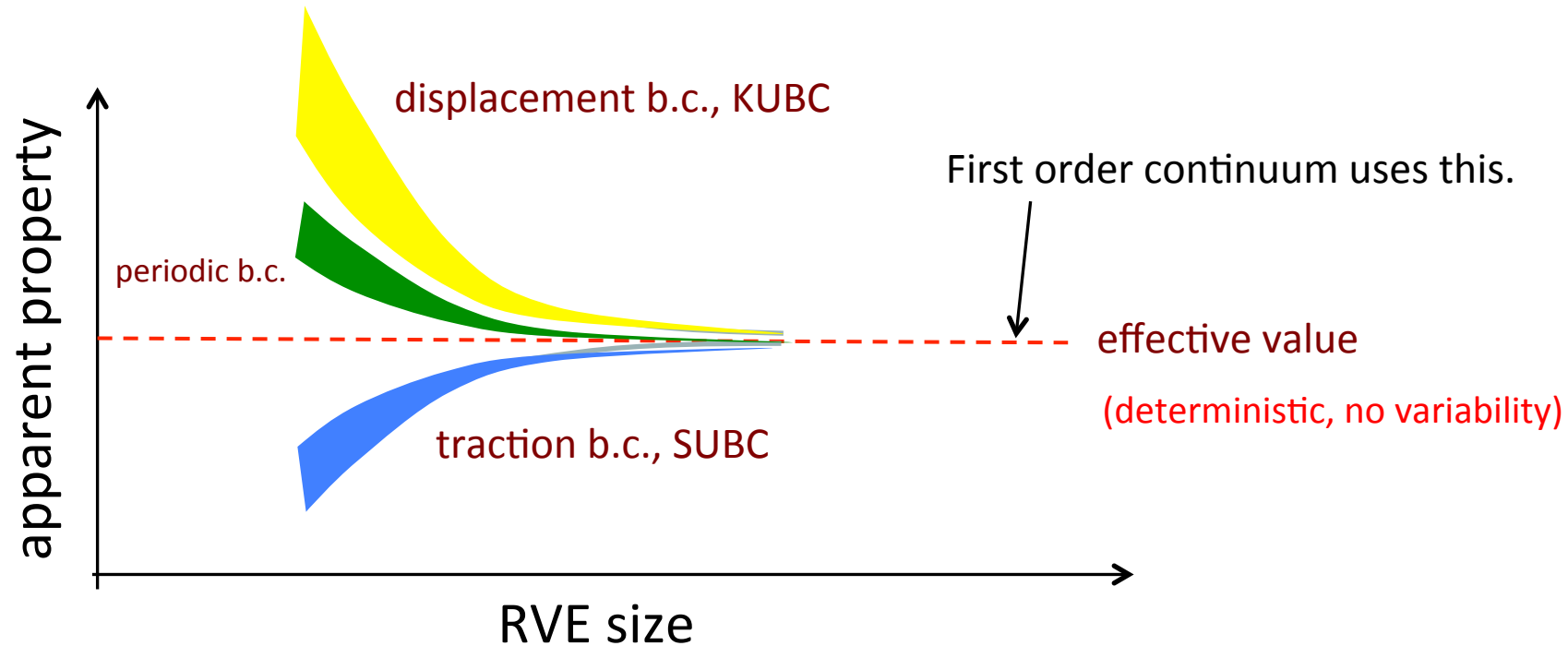
This equivalence is also satisfied energetically:  $\sigma_{ij} \varepsilon_{ij} = \langle \sigma_{ij}^\epsilon \rangle \langle \varepsilon_{ij}^\epsilon \rangle$

Constitutive models map average strain to average stress:

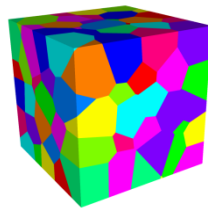
$$\varepsilon_{ij} = \langle \varepsilon_{ij}^\epsilon \rangle \longrightarrow \sigma_{ij} = \langle \sigma_{ij}^\epsilon \rangle$$

# Apparent vs. Effective Material Properties

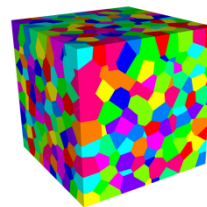
Huet, C. (1990)



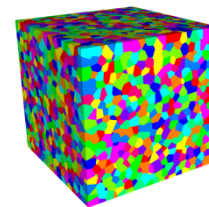
$\varepsilon = 0.32$



$\varepsilon = 0.16$



$\varepsilon = 0.08$



$\varepsilon = 0.04$

# Apparent vs. Effective Material Properties

Huet, C. (1990). "Application of variational concepts to size effects in elastic heterogeneous bodies." *Journal of the Mechanics and Physics of Solids*, 38(6): 813-841.

$C$  = stiffness tensor

finite RVE, **apparent**

infinite RVE, **effective**

$$C_{\sigma}^{\text{app}}(\omega) \leq C \leq C_{\varepsilon}^{\text{app}}(\omega)$$

SUBC

stochastic

deterministic

KUBC

stochastic

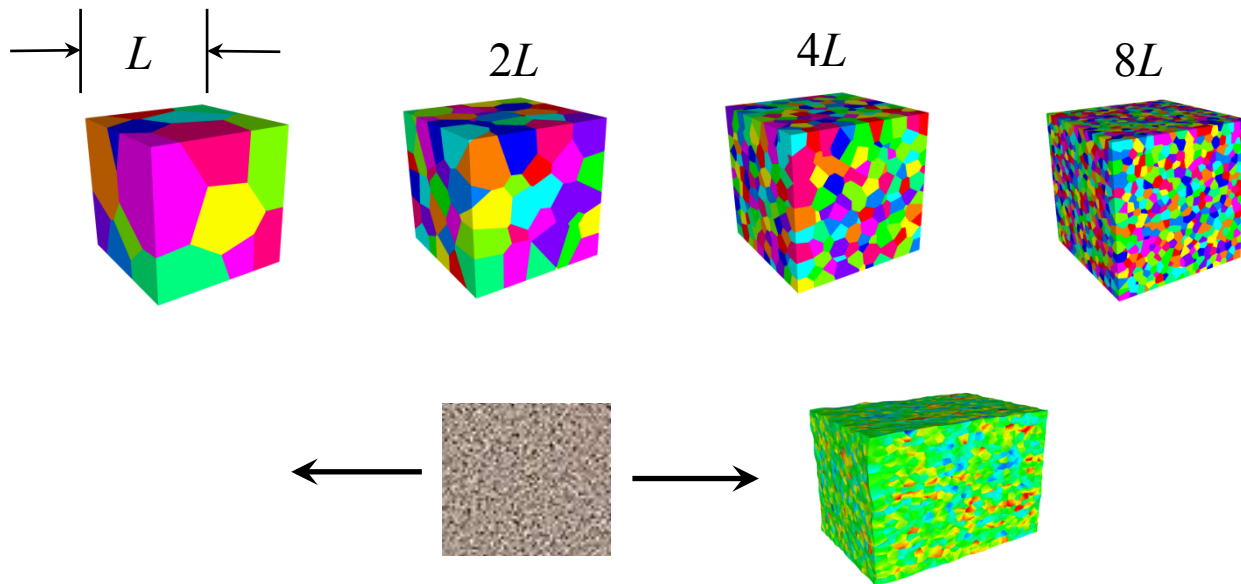
partial ordering defined in an energetic sense:

$$B < A \quad \text{iff} \quad \varepsilon : (A - B) : \varepsilon > 0 \quad \text{for all} \quad \varepsilon \neq 0$$

# Apparent vs. Effective Material Properties

Huet, C. (1990). "Application of variational concepts to size effects in elastic heterogeneous bodies." *Journal of the Mechanics and Physics of Solids*, 38(6): 813-841.

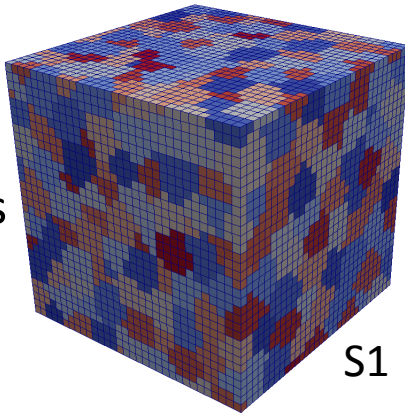
$$C_{\sigma,L}^{\text{app}} \leq C_{\sigma,2L}^{\text{app}} \leq C_{\sigma,4L}^{\text{app}} \leq \dots \leq C_{\sigma,\infty}^{\text{app}} = C$$



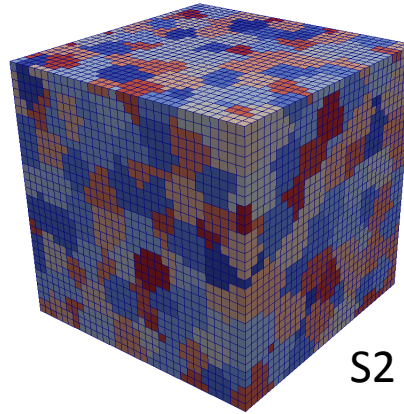


# Stochastic Volume Elements

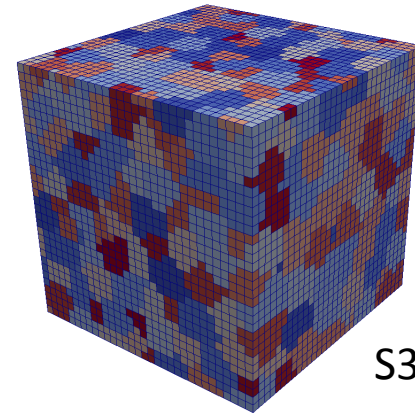
$\sim 8^3$  grains



S1



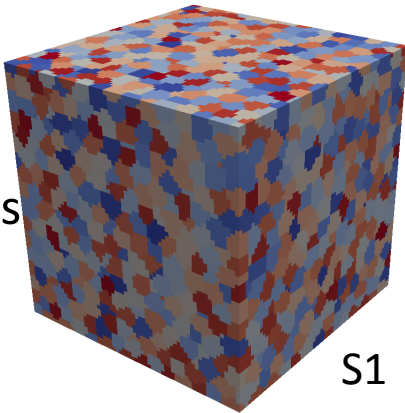
S2



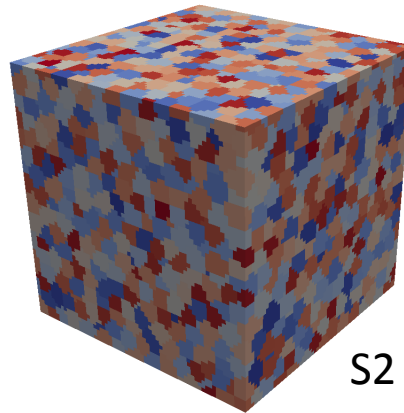
S3

... S100

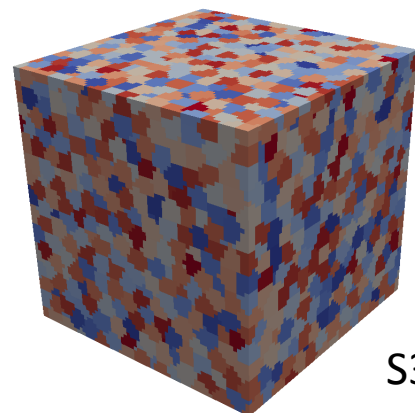
$\sim 16^3$  grains



S1



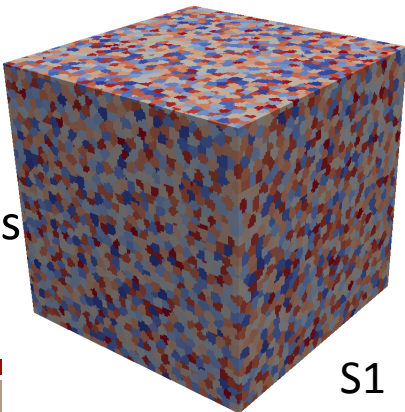
S2



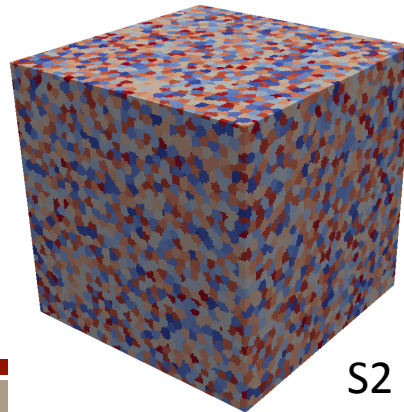
S3

... S100

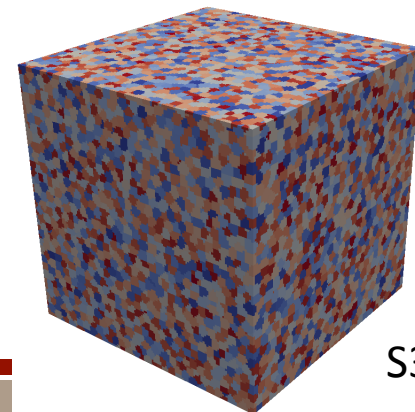
$\sim 32^3$  grains



S1



S2

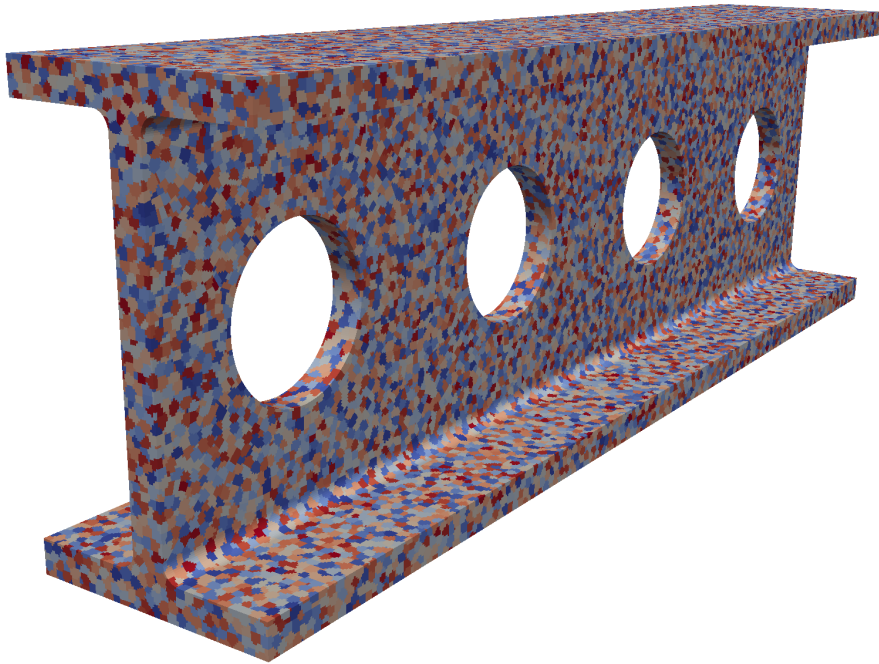


S3

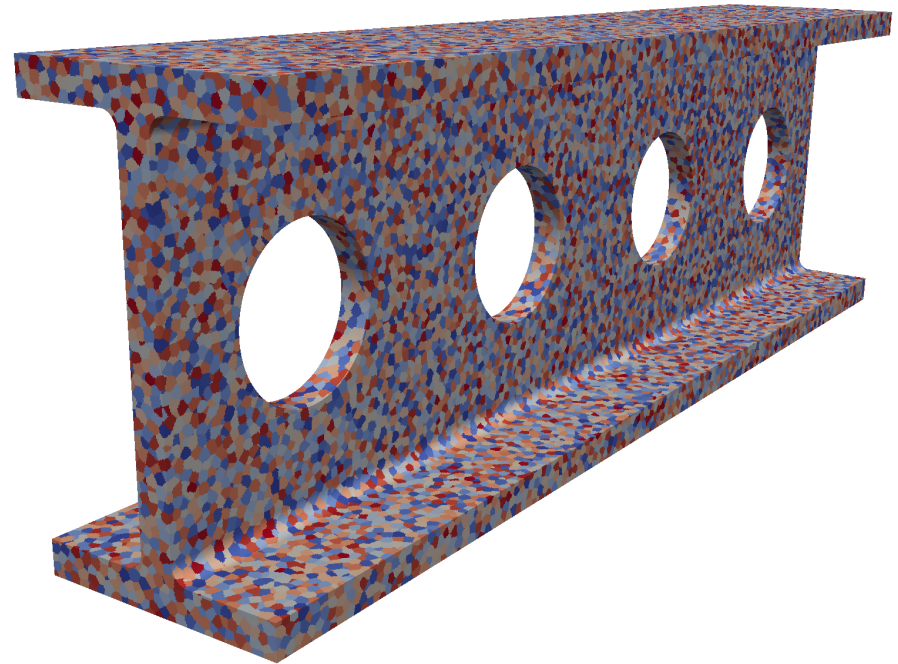
... S100

# Effect of Mesh Refinement

thickness/grain ratio = 4



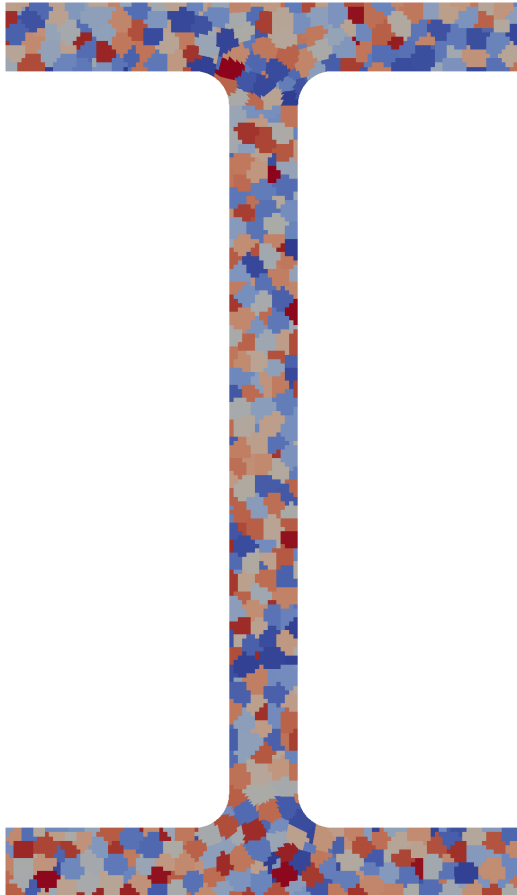
mesh refinement **R3**



mesh refinement **R4**

# Effect of Mesh Refinement

thickness/grain ratio = 4



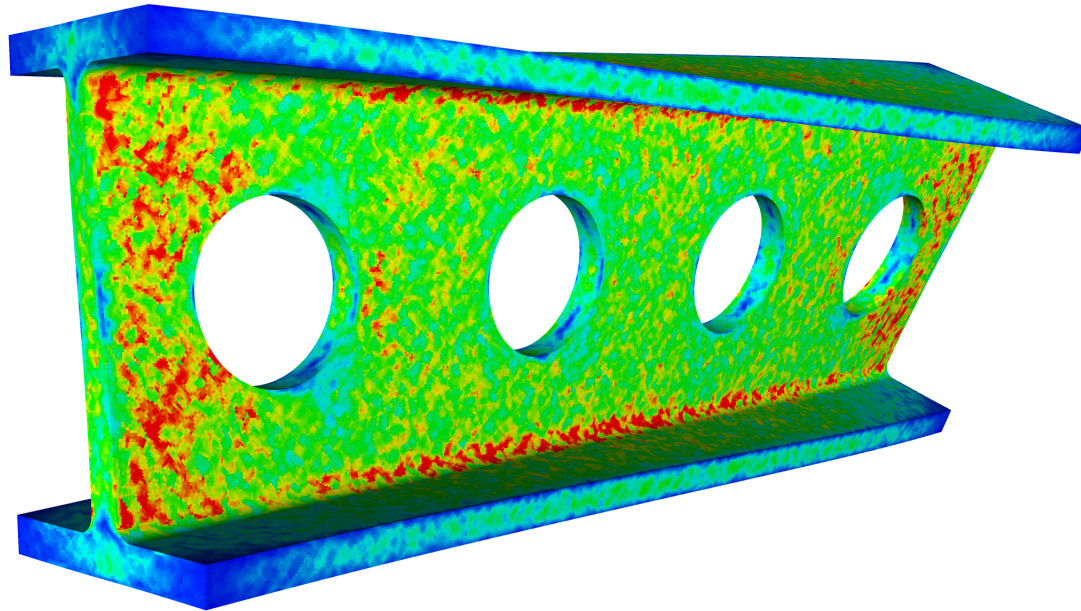
mesh refinement **R3**



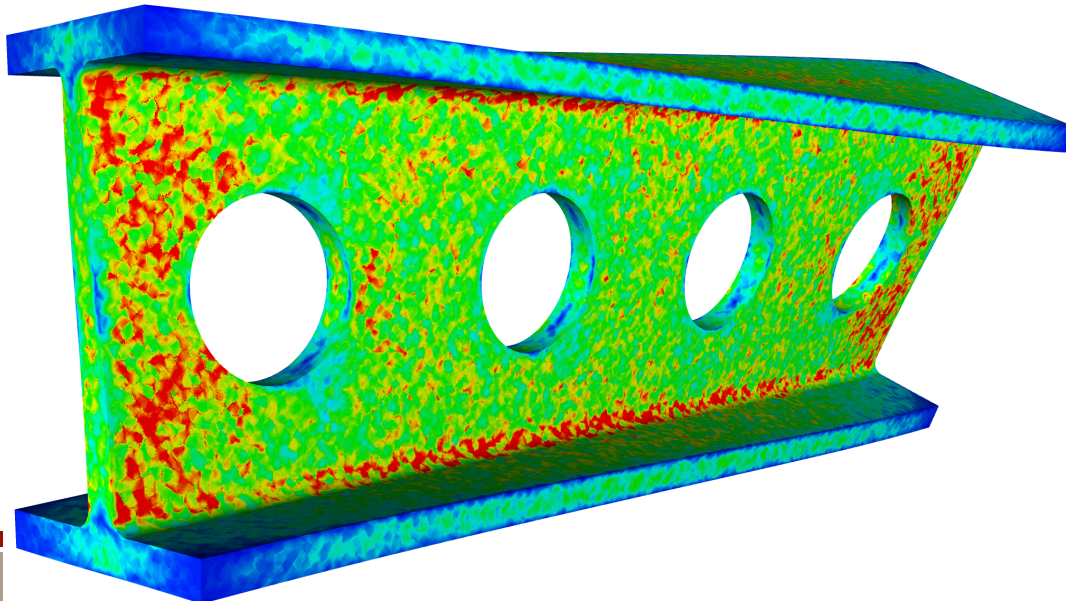
mesh refinement **R4**

# Effect of Mesh Refinement

thickness/grain ratio = 4



mesh refinement **R3**



mesh refinement **R4**



# Effect of Mesh Refinement

thickness/grain ratio = 4

