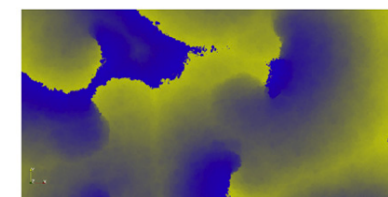
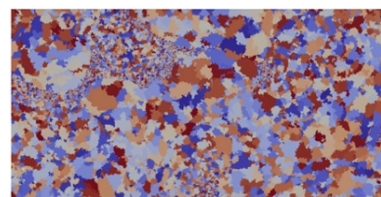
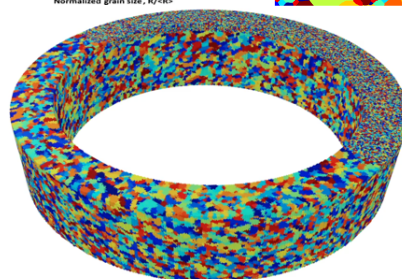
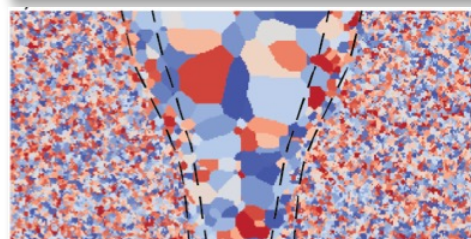
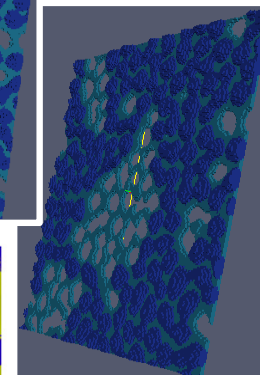
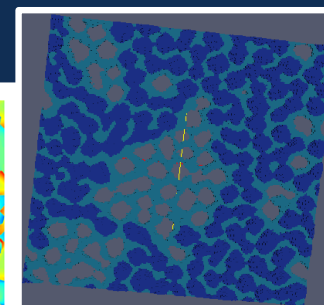
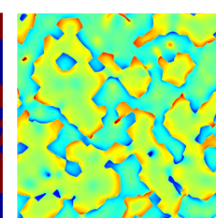
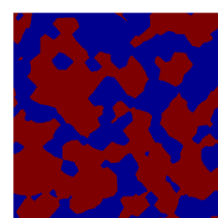
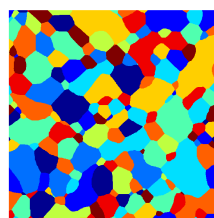
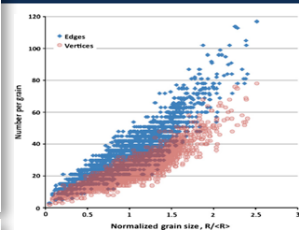
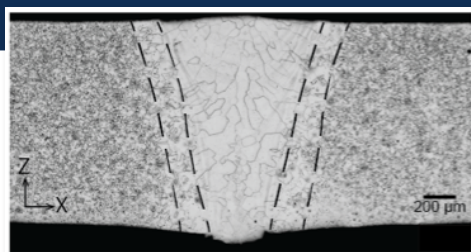


*Exceptional service in the national interest*



# SYNTHETIC MICROSTRUCTURES FOR ADDITIVE MANUFACTURING: GENERATION, QUANTIFICATION & PARTNERING EFFORTS

**Jonathan Madison**, Theron Rodgers, Kyle Johnson, Joe Bishop, Veena Tikare

<sup>1</sup>Sandia National Laboratories, Albuquerque, NM



Sandia National Laboratories is a multi-mission laboratory managed and operated by National Technology and Engineering Solutions of Sandia, LLC., a wholly owned subsidiary of Honeywell International, Inc., for the U.S. Department of Energy's National Nuclear Security Administration under contract DE-NA0003525. SAND2017-XXXX

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- **Joe Michael**
- **Ben Reedlunn**
- **Joe Bishop**
- **Kyle Johnson**



**PennState**

- **Todd Palmer et al., Pennsylvania State University**  
*Center for Innovative Materials Processing through Direct Digital Deposition (CIMP-3D)*

- **Surya Kaladindi et al., Georgia Institute of Technology**



- **Amy Clark et al., Colorado School of Mines**



- **Joe McKeown et al., Lawrence Livermore National Laboratories**





# Outline

- I. A DATABASE APPROACH TO ADDITIVE
- II. COMPUTATIONAL LENGTHSCALES & TOOLS
- III. SPARKS FRAMEWORK & APPROXIMATING SOLIDIFICATION
- IV. ADDITIVE CASE STUDIES
- IV. PARTNERING EFFORTS
  - *Thin Films*
  - *Data Science*
  - *Thermal History Models*
  - *Direct Numerical Simulation for Mechanics*
- V. CONCLUDING THOUGHTS & SUMMARY

# GRANTA's Approach to Additive

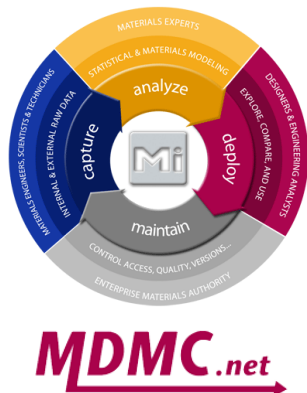


## Leverage core expertise of MDMC members

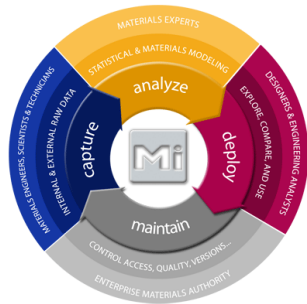
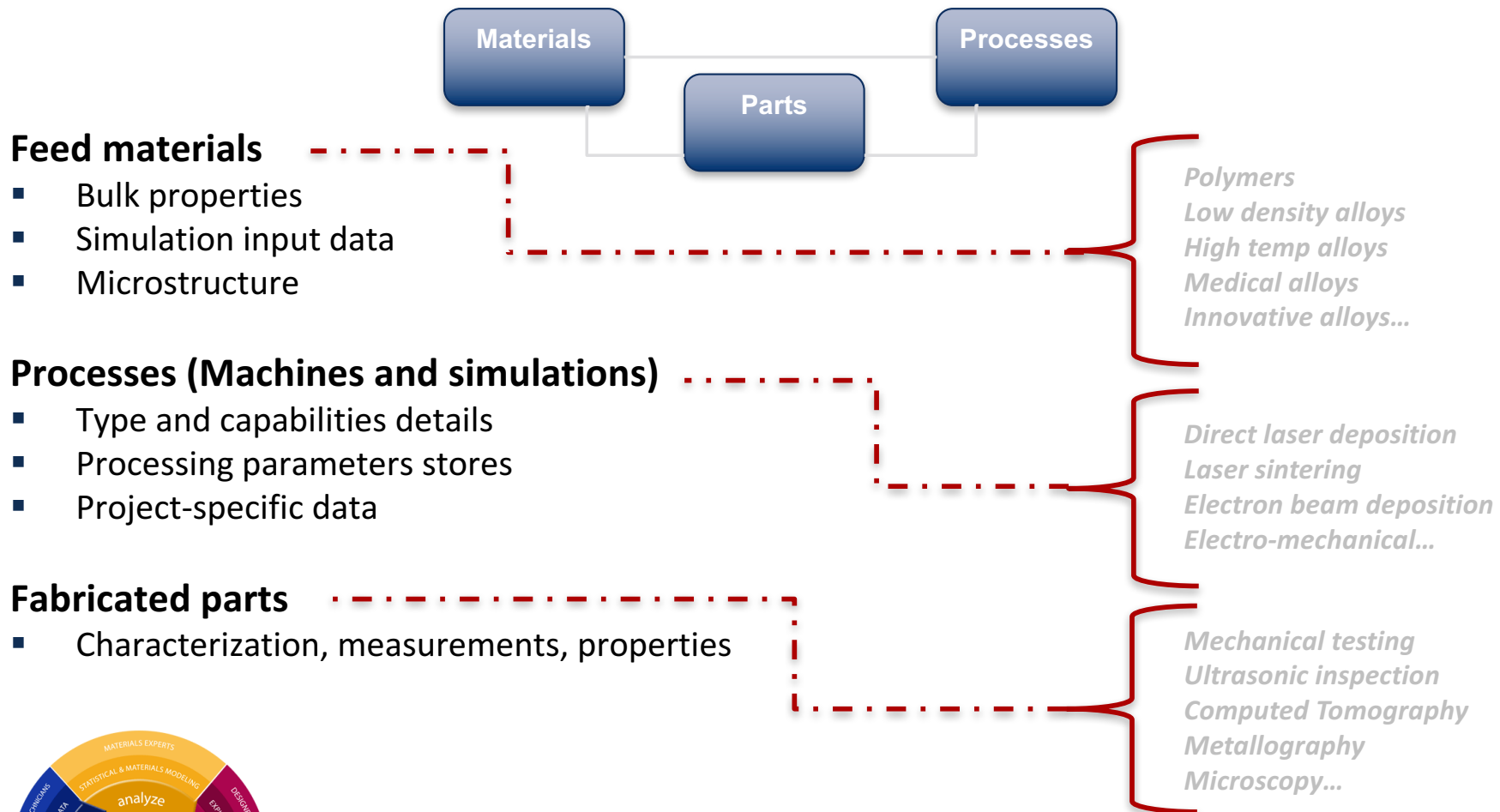
- Materials, manufacturing, data management
- Software infrastructure and networks

## Focus

- Schema development, software & network integration,
- standards, interfacing with AM initiatives worldwide



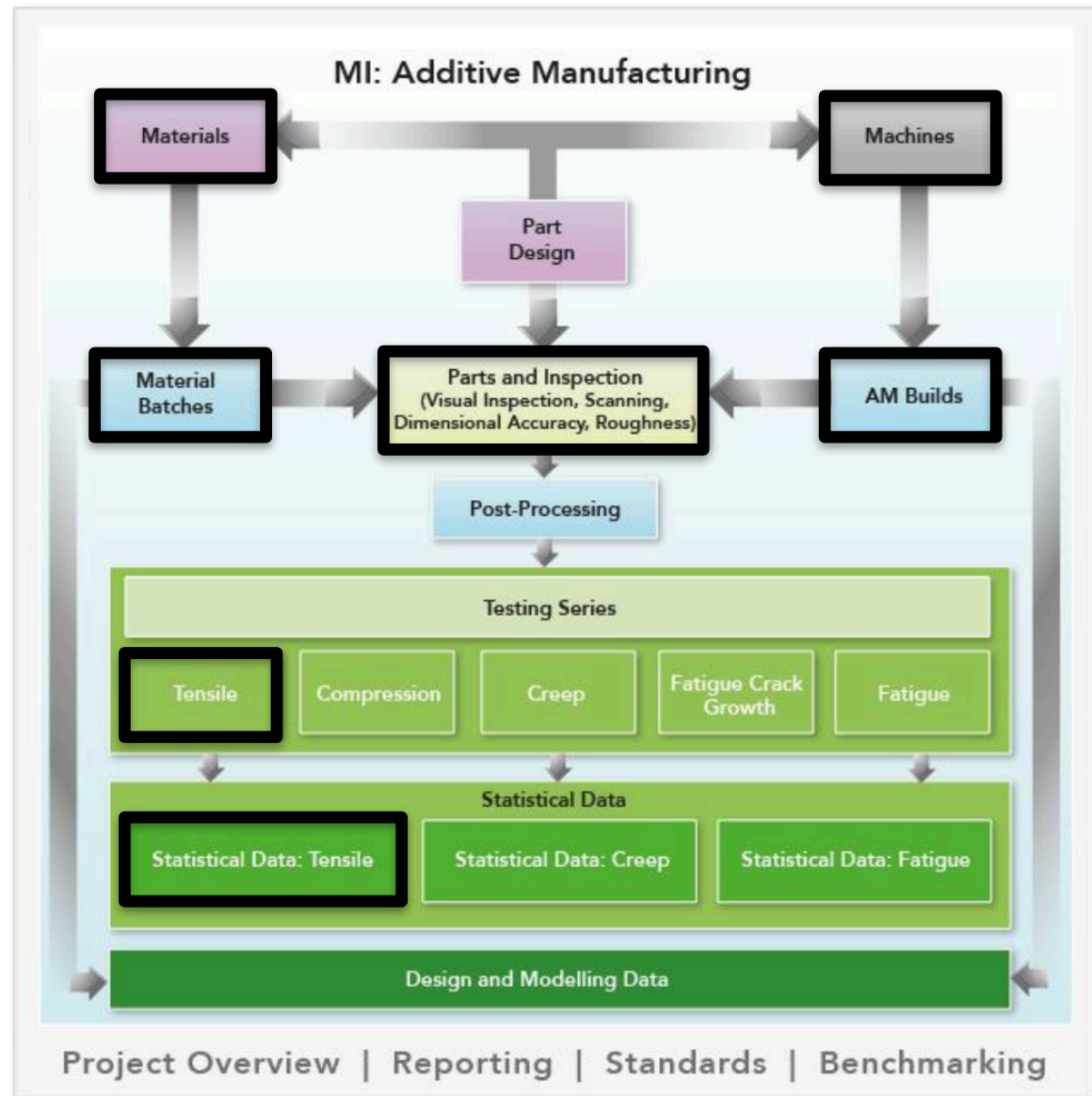
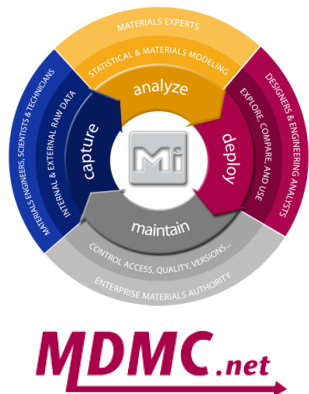
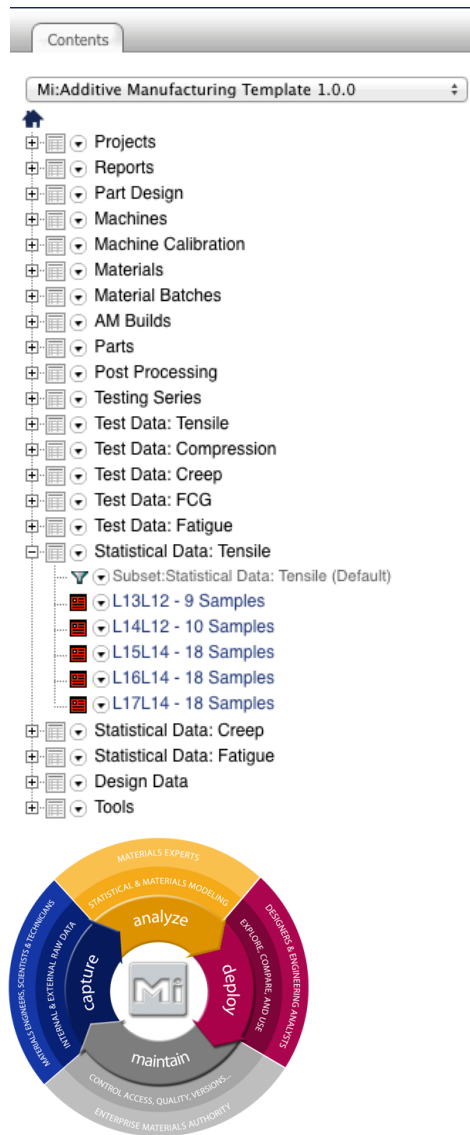
# AM Schema Requirements



MDMC.net

GRANTA  
MATERIAL INTELLIGENCE

# AM Schema Visualized





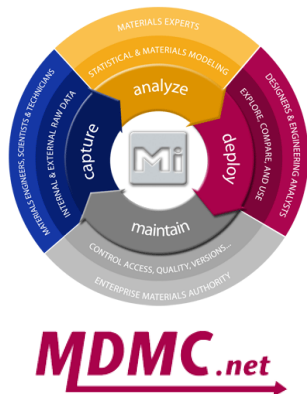
# AM Schema Attributes Count

## Granta MI, AM Schema Layout Headings

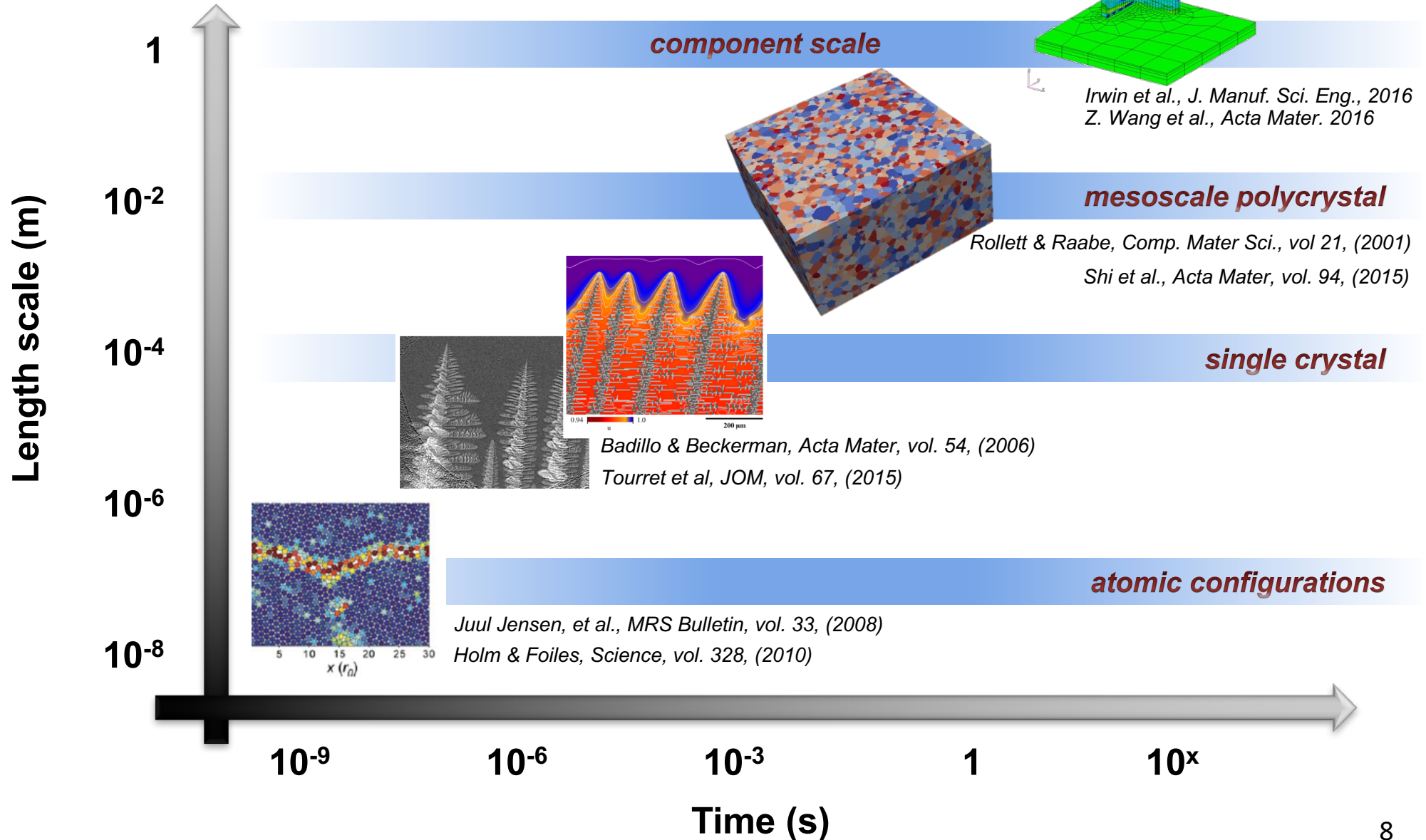
Layout Headings - Granta AM Schema v1.01					
Machines	Materials	Part Design	Material Batches	Builds	Parts
General Information	General Information	General Information	Project Information	Project Information	Project Information
Calibration	General Properties	Original Design	General Batch Information	General Information	Part Information
Machine Specifications	Composition overview	Re-Design	Manufacturing	Build Information	Part Specifications
Material	Bulk Mechanical Properties	Dimensions	Material Quality	General Build Parameters	Samples
Machine Properties	Bulk Thermal Properties	General Material Properties	Particle Properties and Size Distribution	Build Atmosphere	Visual Inspection
Build Environment	Bulk Electrical Properties	Processing	Interstitial contamination	Material Used	Accuracy Testing
Laser Properties	Biological	Static Tensile Properties	Flowability	Support	NDT Testing
Electron Beam Properties	Chemical	High-cycle fatigue properties	Wire Properties	Filament Information	Post Processing
	Eco	Fracture Toughness	Chemical Analysis and Composition	Substrate	Heat Treatment
	Cost	Fatigue Crack Growth		Quality of Welding Consumables	HIP
	Safety and Handling	Surface Roughness Requested		Build Alarms	Machining
	General Information	Other Requested Properties		Themes Used	Laser Polishing
	Requirements	Final Part Details		Powder Build Parameters	Other Post Processing
	Composition	Quality Assurance		Wire Build Parameters	
	Physical Properties	Key Benefits		Laser Properties	
	Further Information			Electron Beam Properties	
				Arc Properties	
				In-Process Rolling	
				In-Process Analysis	

Currently Includes **800+** individual attributes

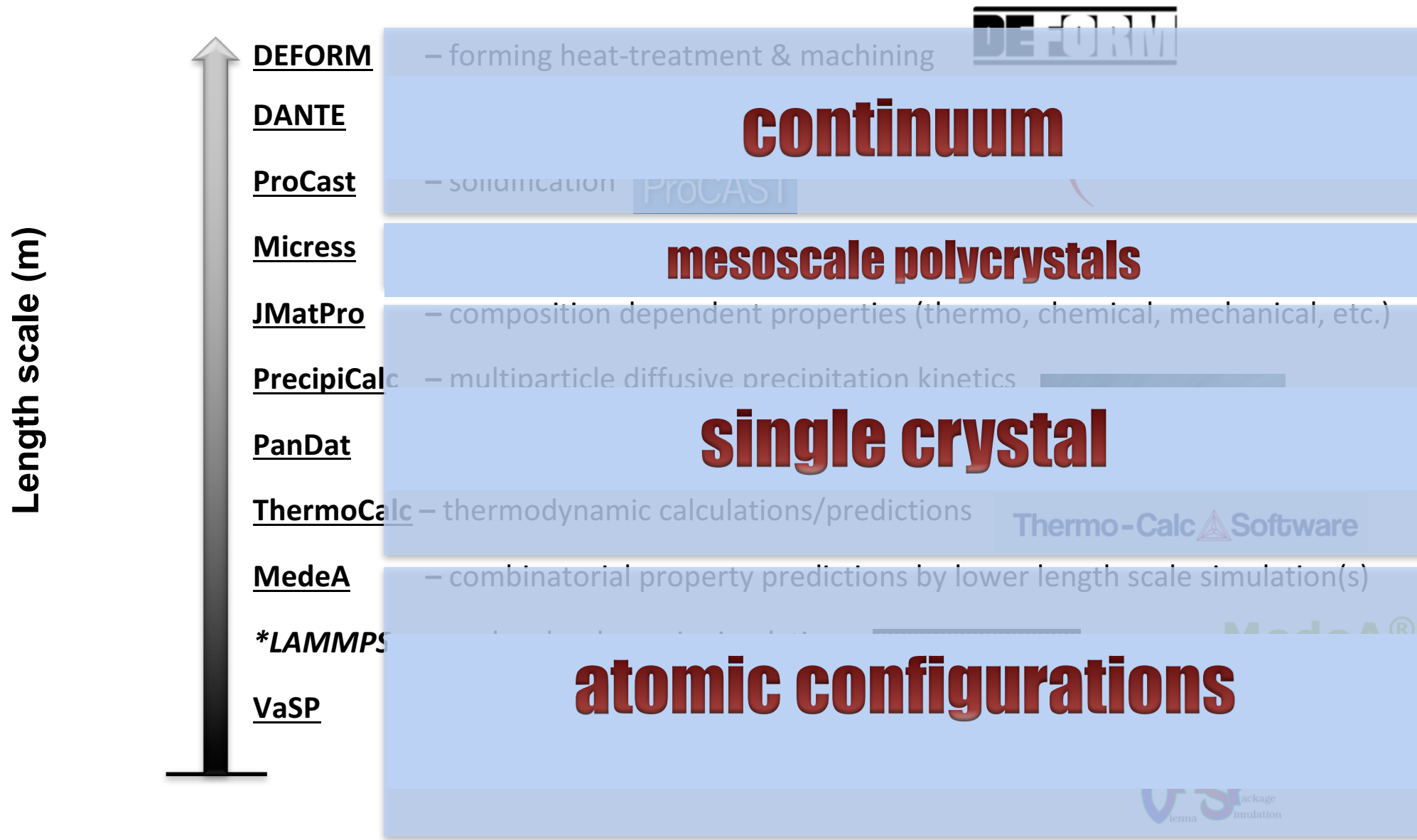
Traditional brute force design of experiment approaches are no longer tenable. Efficient & flexible modeling approaches in additive are needed now more than ever.



# Computational Length Scales



# Computational Tools

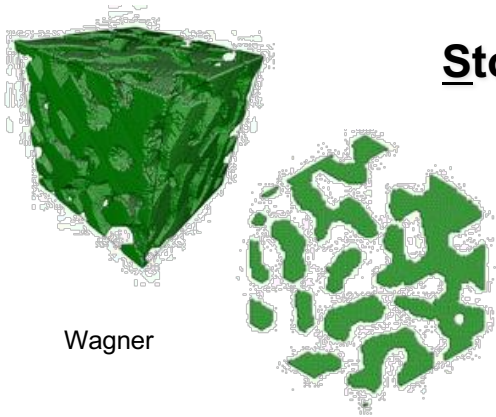


# Sim Framework

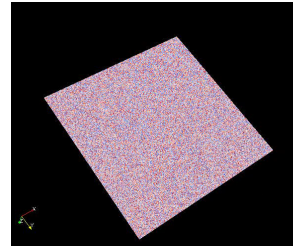
## SPPARKS

Stochastic Parallel PARTicle Kinetic Simulator

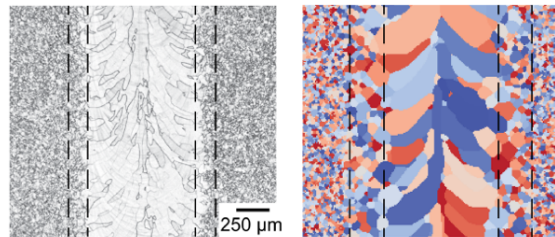
<http://spparks.sandia.gov>



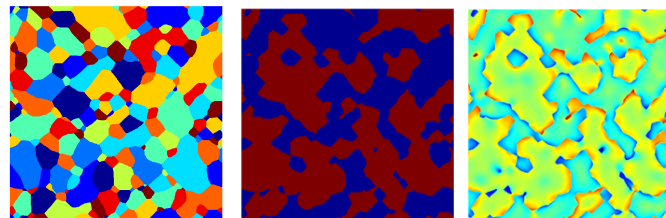
Wagner



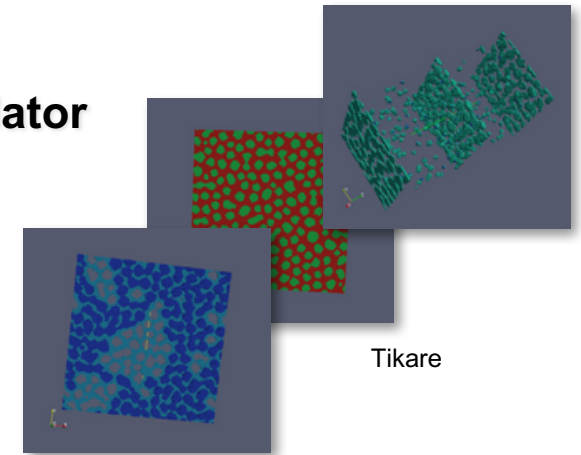
Holm & Battaile, *JOM* (2001) vol. 53, pp. 20-23



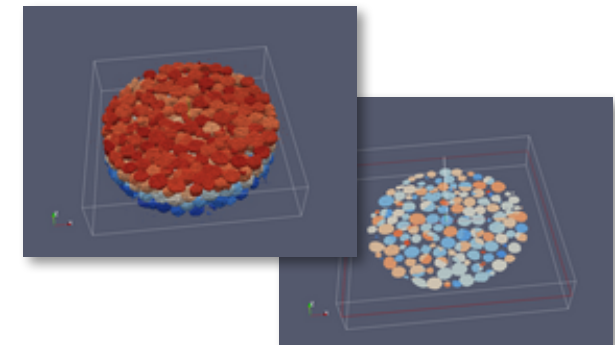
Rodgers, Madison, Tikare & Maguire, *JOM* (2016) vol. 68, pp. 1419-1426



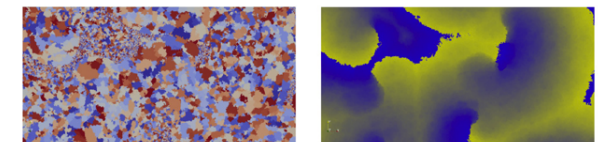
Homer, Tikare & Holm, *Comp. Mat. Sci.* (2013) vol. 69 pp. 414-423



Tikare



Cardona et al. *J. Am. Ceram. Soc.* (2012) pp. 1-12



Madison, Tikare & Holm, *J. Nuc. Mat.* (2012) vol. 425 pp. 173-180

- Open source kinetic Monte Carlo platform with user-editable “apps” for specific applications
- Used to study several mesoscale phenomena including sintering, recrystallization, vacancy diffusion, grain growth and welding
- Problems are easily parallelized and a range of Monte Carlo solvers are available



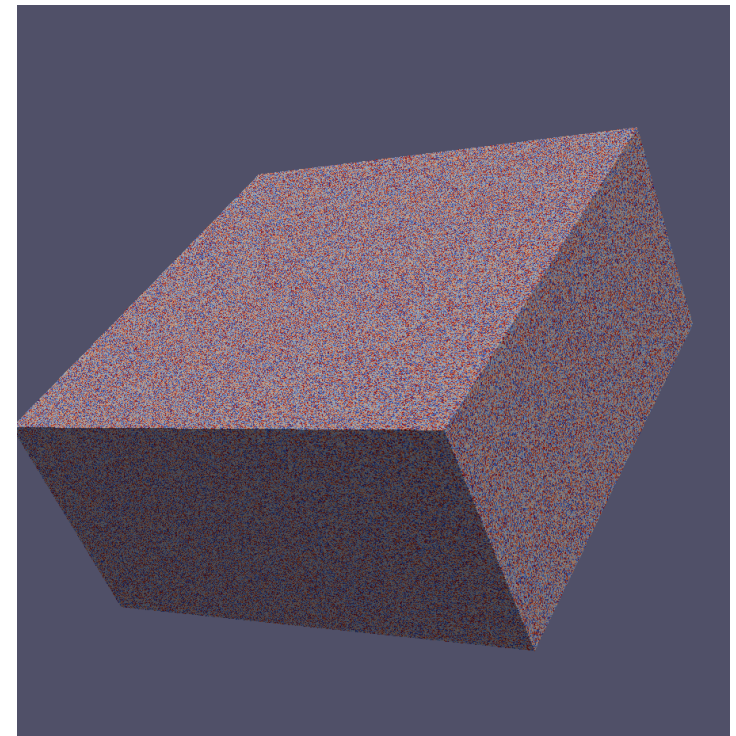
# Potts kinetic Monte Carlo

- The Potts model is widely used to simulate curvature-driven grain growth
- The simulation domain is divided into a discrete lattice and each site is assigned a grain ID or “spin”.
- Grain boundary motion occurs by changing a site’s grain ID to that of a neighboring lattice site and determining the resulting change in system energy.
- The change is accepted/rejected via the
  - Metropolis function:

$$P = \begin{cases} \exp\left(\frac{-\Delta E}{k_B T_s}\right) & \text{if } \Delta E > 0 \\ 1 & \text{if } \Delta E \leq 0 \end{cases}$$

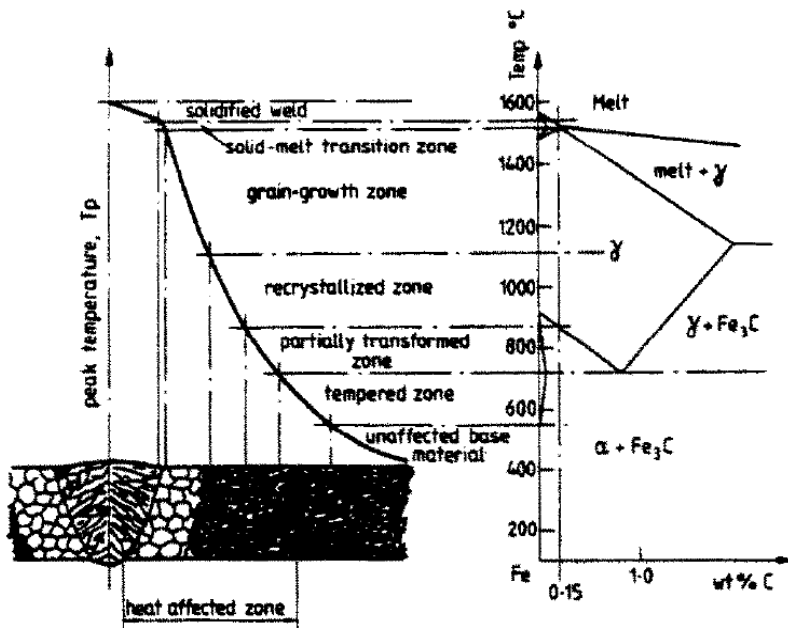
- where system energy, E, is determined by:

$$E = \frac{1}{2} \sum_{i=1}^N \sum_{j=1}^{26} (1 - \delta(q_i, q_j))$$

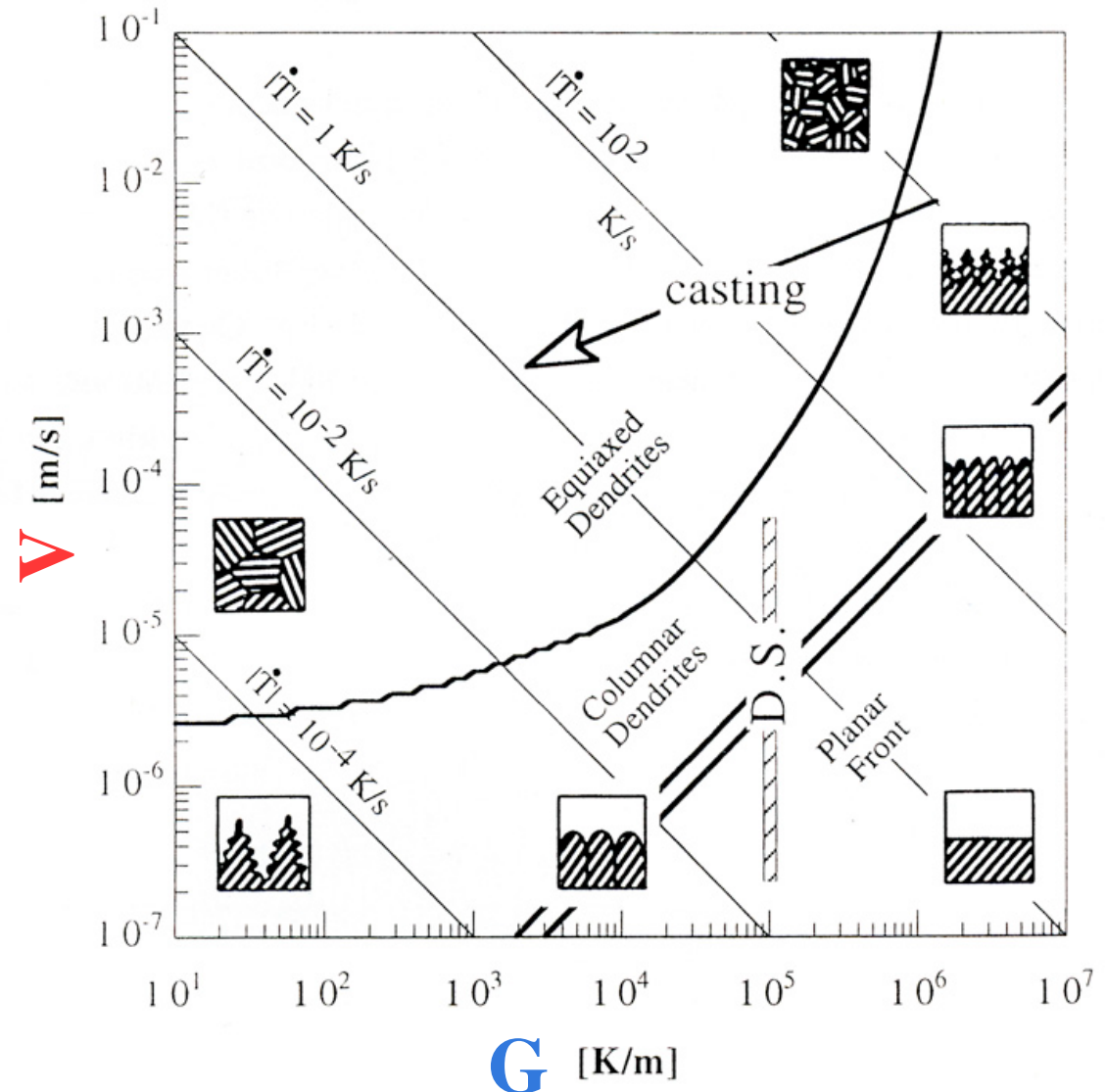


# Solidification

- In directional solidification, it has been widely understood that select combinations of **thermal gradient ( $G$ )** and **solidification front velocity ( $V$ )** produce specific grain morphologies in single phase metallic systems



M.F. Ashby, K.E. Easterling, *Acta Metall.*,  
(1982) vol. 30, pp. 1969-1978

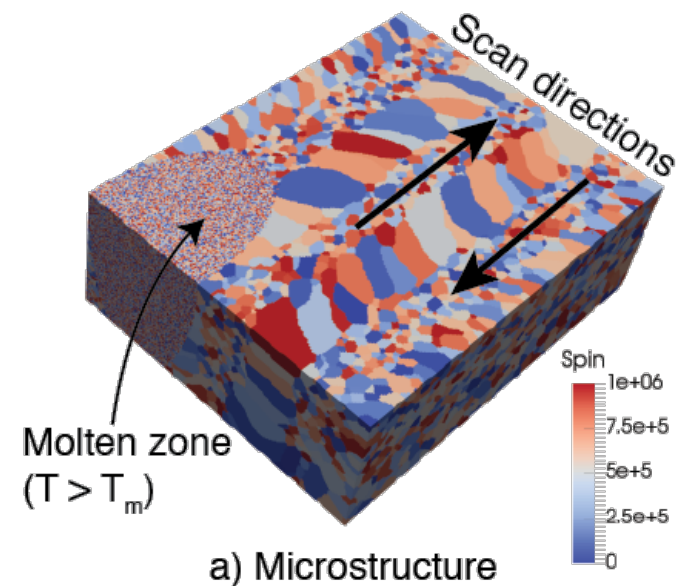
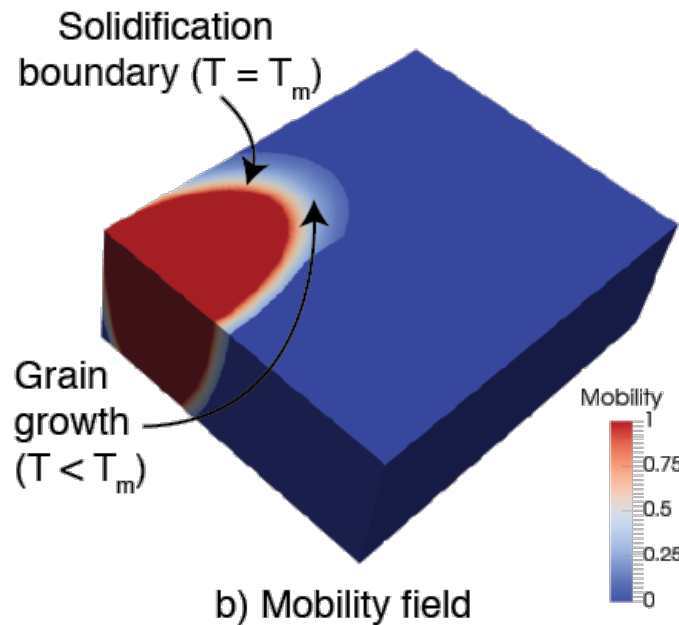
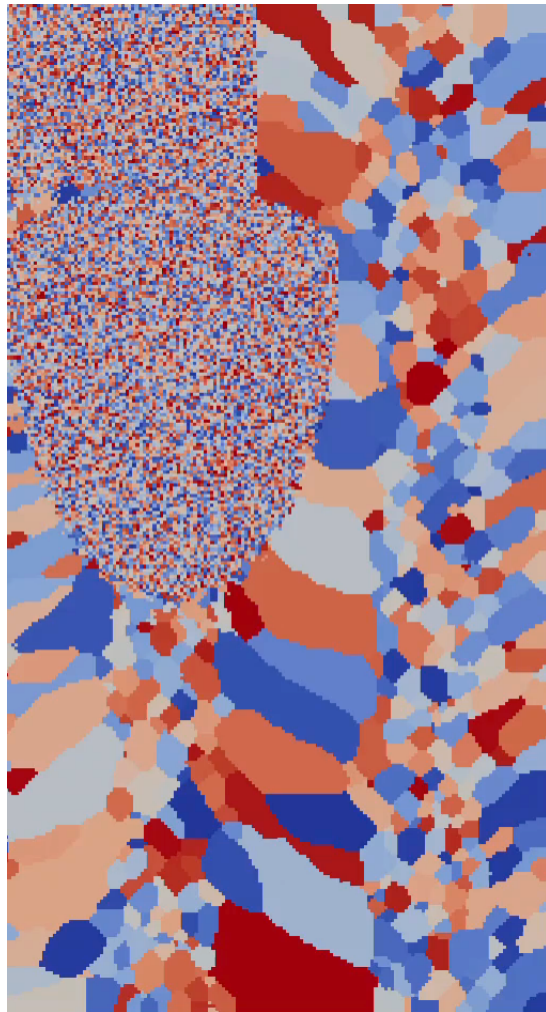


W. Kurz, & D.J. Fisher, *Fundamentals of Solidification*, Chapter 4:  
Solidification Microstructure, Trans Tech Publications (Enfield, NH, 1998)

# kMC & Solidification

$$M(T) = M_o \exp\left(\frac{-Q}{RT}\right)$$

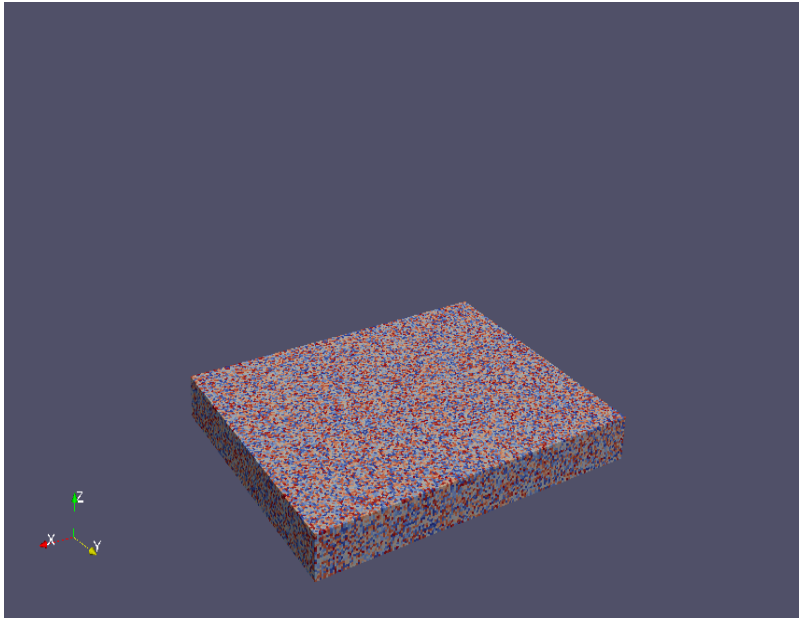
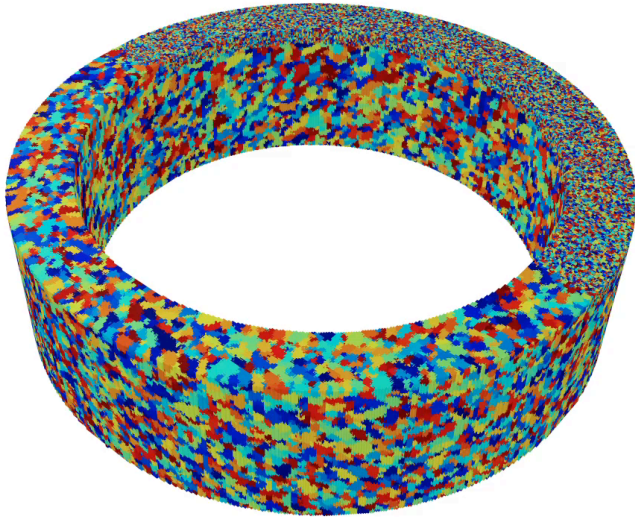
$$P = \begin{cases} M(T) \exp\left(\frac{-\Delta E}{k_B T_s}\right), & \text{if } \Delta E > 0 \\ M(T), & \text{if } \Delta E \leq 0 \end{cases}$$



- The molten zone randomizes grain identities when it enters a region.
- Along the trailing surface, voxels either join existing columnar grains or form new grains.
- The temperature gradient creates a corresponding gradient of grain boundary mobilities via an Arrhenius relationship.

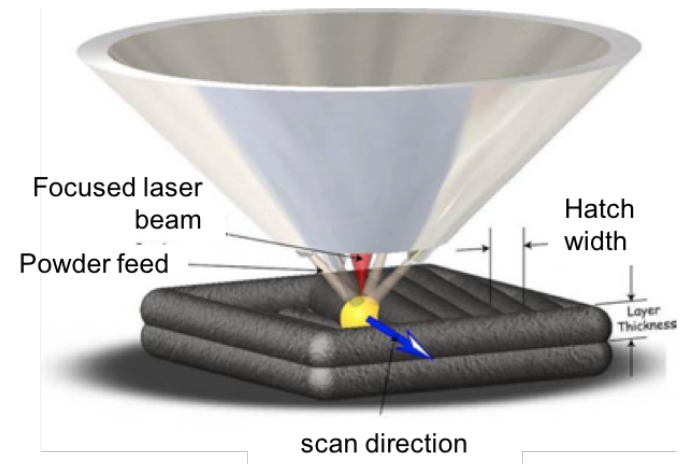


# Additive Processes

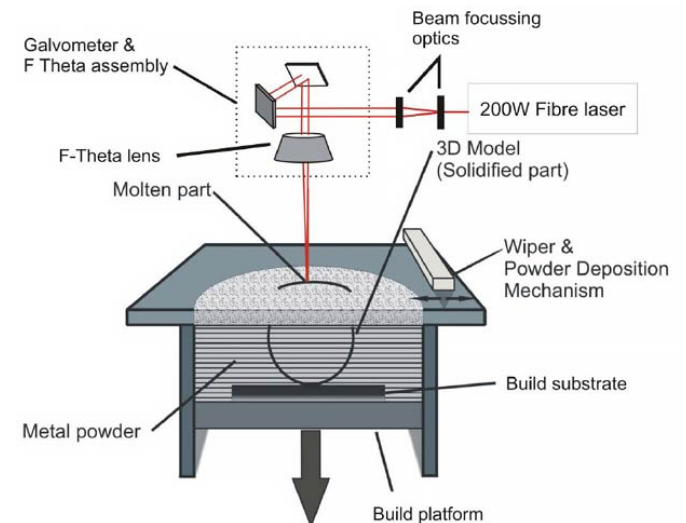


modified Kinetic Monte Carlo (MkMC) via SPPARKS

## LASER ENGINEERED NET SHAPING



## POWDER BED FUSION



<http://www.fusionimplants.com>



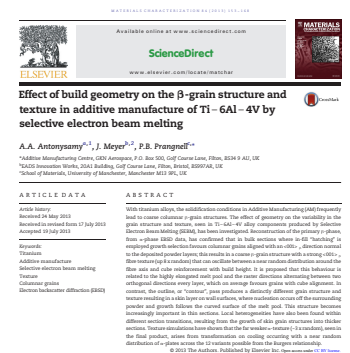
- Thin-wall builds of Inconel 718 by DED

Parimi et al., *Mater. Charact.* (2014), vol. 89, pp. 102-111



- Two-pass build of Ti-6V-4Al by SEBM

A. Antonysamy et al., *Mater. Charact.* (2013), vol. 84, pp. 153-168



EPJ Web of Conferences **94**, 01001 (2015)  
DOI: 10.1051/epjconf/20159401001  
© Owned by the authors, published by EDP Sciences, 2015

- High-power LENS build of 304L Stainless Steel

Nishida et al., Proc. of DYMAT 2015

## Dynamic compressive response of wrought and additive manufactured 304L stainless steels

<sup>1</sup> Sandia National Laboratories, 1515 Eubank Blvd. SE, New Mexico, USA  
<sup>2</sup> Pennsylvania State University, MatSE Department, 301 Applied Science Building, Pennsylvania, USA

**Abstract.** Additive manufacturing (AM) technology has been developed to fabricate metal components that include complex geometries, prototype fabrication, small lot production, precision repair or feature addition, and tooling. However, the mechanical response of the AM materials is a concern to meet requirements for specific applications. Differences between AM materials as compared to wrought materials might be expected, due to possible differences in porosity (voids), grain size, and residual stress levels. When these differences are not desired, materials are designed to be different from wrought materials. In this study, the mechanical response of AM tooling to be fully characterized and understood for reliable designs. In this study, a 304L stainless steel was manufactured with AM technology. For comparison purposes, both the AM and wrought 304L stainless steels were dynamically characterized in compression. Kinking behavior and techniques. Their dynamic compressive stress-strain curves were obtained and the strain rate effects were investigated. Both the AM and wrought 304L stainless steels were tested in compression at strain rates of 1000 to 10,000 s<sup>-1</sup>. The dynamic compressive response between the AM and wrought 304L stainless steels was performed. SAND2015-0993-C.

## 1. Introduction

Additive manufacturing (AM), also commonly known as 3D printing, has been increasingly utilized in many applications that include rapid prototyping and manufacturing as well as mass customization for automobile, aerospace construction, and defense industries. For example, Sandia National Laboratories developed a new technology called Laser Shaping (LENS) utilizing a type of AM technology which utilizes a combination of metal powders and a laser to create metallic components designed by computer-aided design (CAD) models [1]. New AM processes that are LENS-like have been recently developed to create structures at much lower cost than traditional AM. As AM materials are created, residual stresses may occur and grain structures may change [2], which can lead to uncertainty in the mechanical behaviour of the material. Particularly, the AM materials may be utilized in abnormal mechanical environments in which the material is subjected to cyclic loading. Understanding the dynamic response of the AM materials is critical for applications in terms of reliability

In this study, a wrought and AM processed 304 stainless steel (SS) were dynamically characterized to compare compressive stress-strain curves at various strain rates with a split Hopkinson pressure bar (SHPB) or Kolsky compression bar.

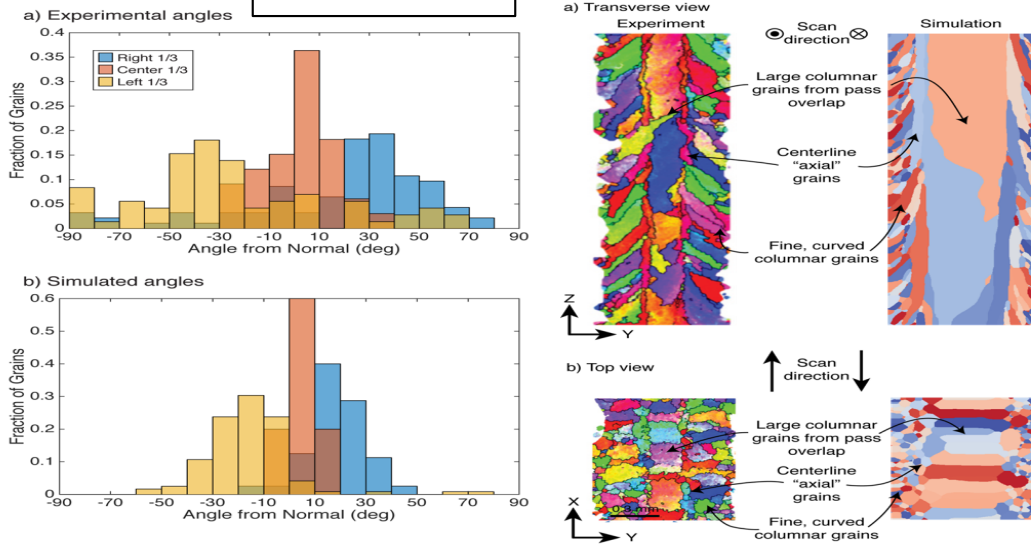
## 2. Materials & experimental procedure

\* Corresponding author: [enishi@u-aadai.gov](mailto:enishi@u-aadai.gov)

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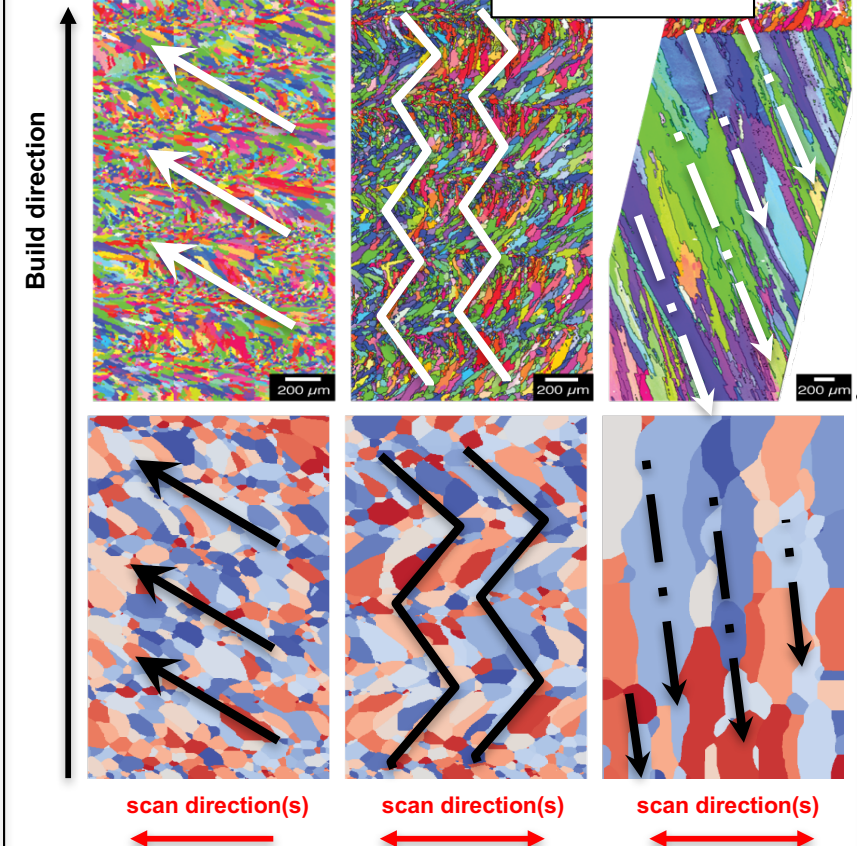
# Additive Case Studies

## Ti-6Al-4V



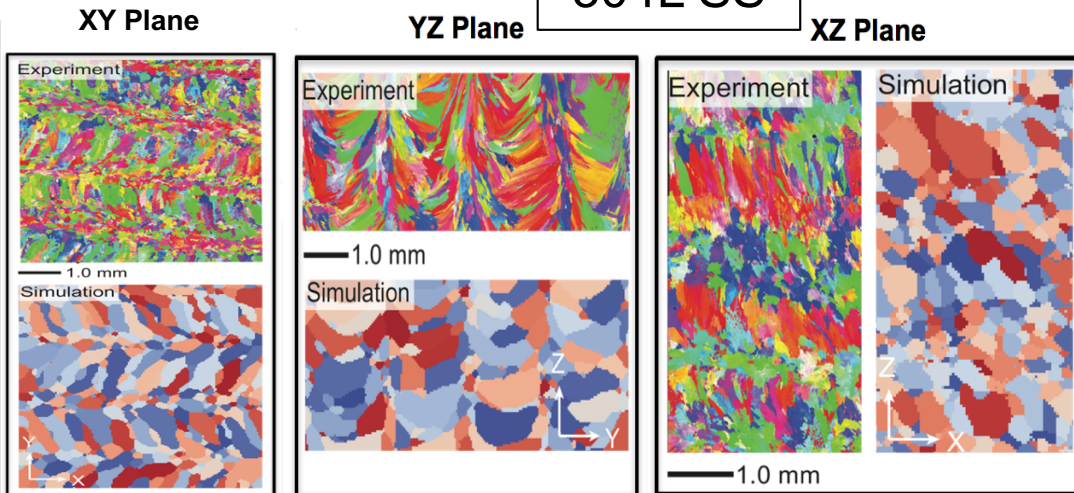
A. Antonysamy et al., *Mater. Charact.* (2013), vol. 84, pp. 153-168

## IN 718



Parimi et al., *Mater. Charact.* (2014), vol. 89, pp. 102-111

## 304L SS



Nishida et al., *Proc. of DYMAT 2015*

Computational Materials Science 135 (2017) 78–89

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journal homepage: [www.elsevier.com/locate/commatsci](http://www.elsevier.com/locate/commatsci)

Simulation of metal additive manufacturing microstructures using kinetic Monte Carlo

Theron M. Rodgers<sup>a,\*</sup>, Jonathan D. Madison<sup>b</sup>, Veena Tikare<sup>c</sup>

<sup>a</sup>Computational Materials & Data Science, Sandia National Laboratories, PO Box 5800 MS-1411, Albuquerque, NM 87185, USA

<sup>b</sup>Materials Mechanics, Sandia National Laboratories, PO Box 5800 MS-0889, Albuquerque, NM 87185, USA

<sup>c</sup>Multiscale Science, Sandia National Laboratories, PO Box 5800 MS-1321, Albuquerque, NM 87185, USA

CrossMark

# **Application to Thin Film Grain Growth**

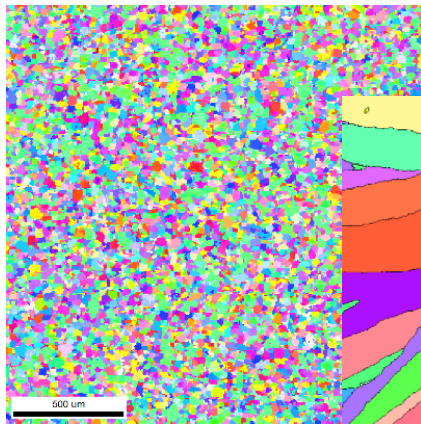


# Grain Growth in Thin Films

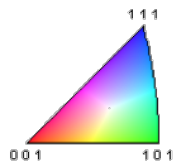
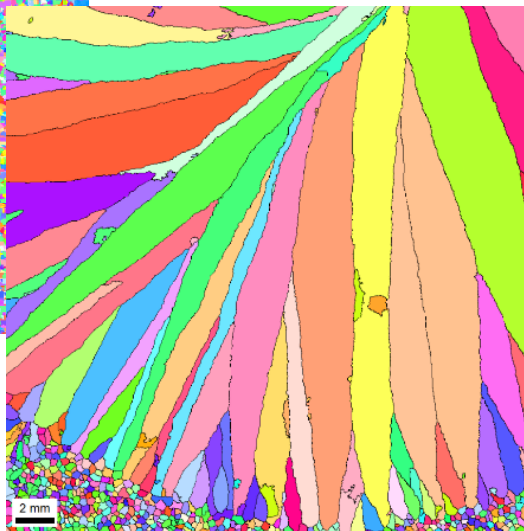
- Thin Films of Al-Si binary alloys melted to undergo rapid solidification while being observed in a time-resolved dynamic TEM (DTEM).
- Solidification interface profiles are extracted and used as inputs for a 2D kMC model.
- Simulations are now being modified to allow incorporation of crystallographic texture

## Experimental ASTAR

Initial

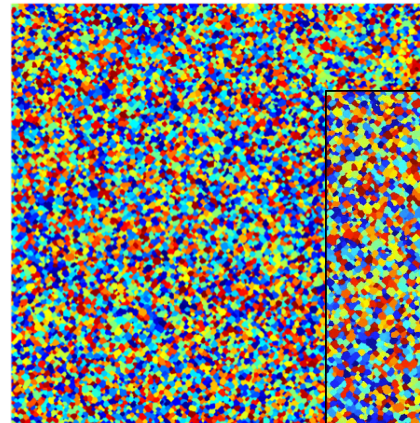


Re-solidified

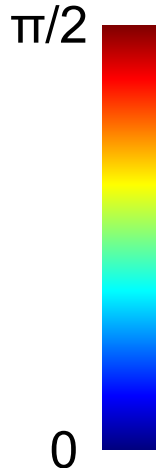
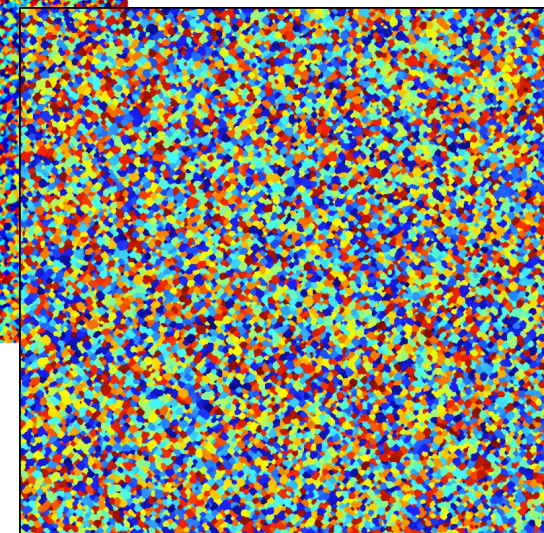


## Simulation

Initial



Re-solidified

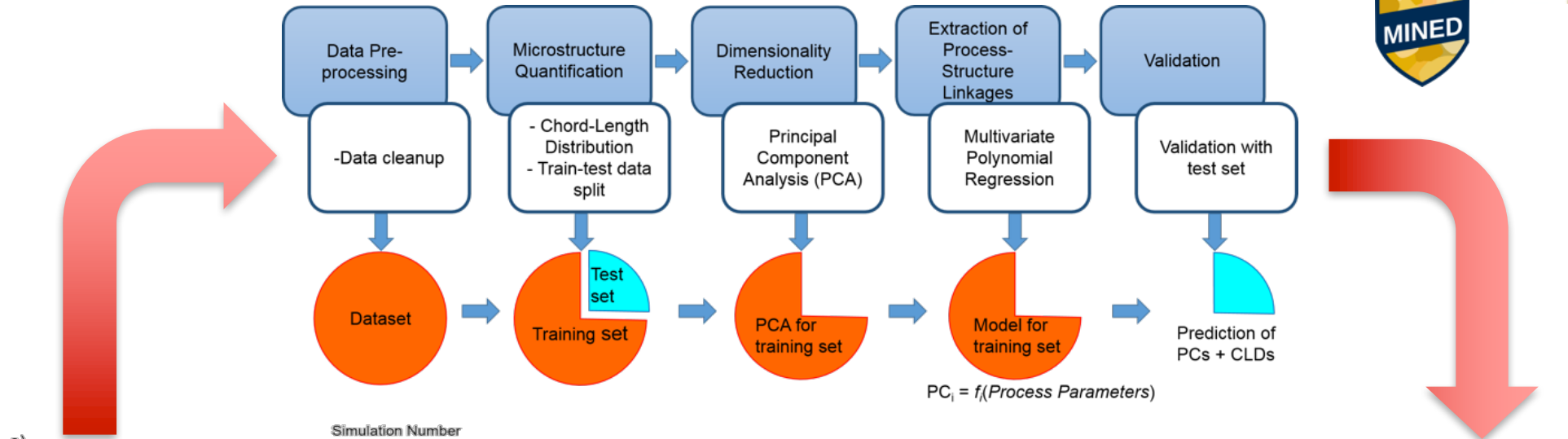




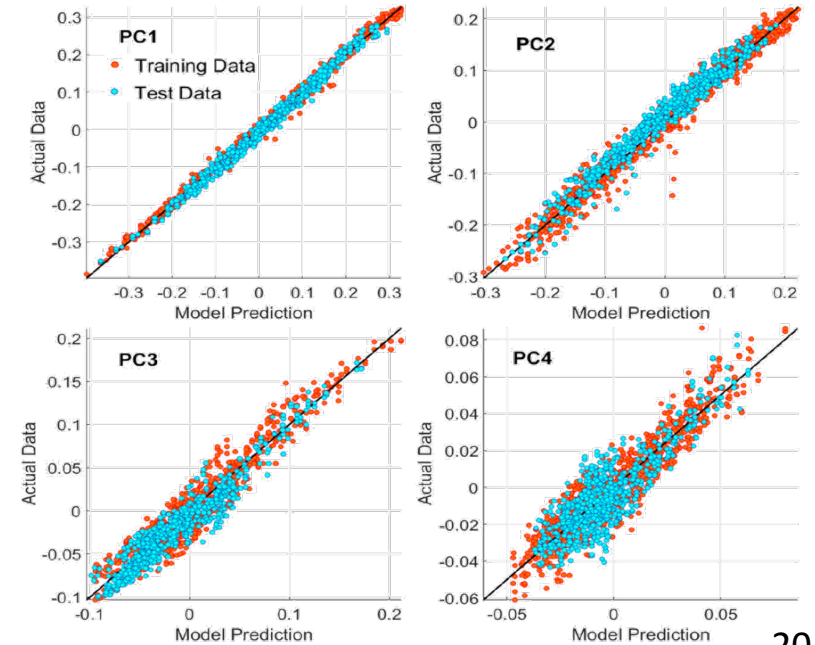
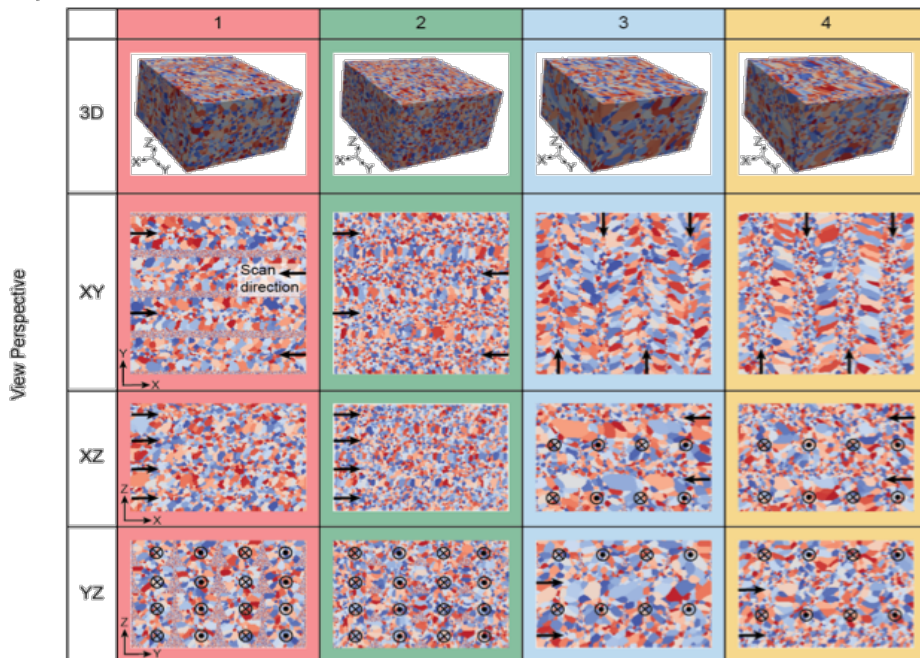
# **Application to Data Science**

# Exploring AM via Data Science I

- Developed a reduced-order model for microstructure prediction using chord length distributions (CLDs).

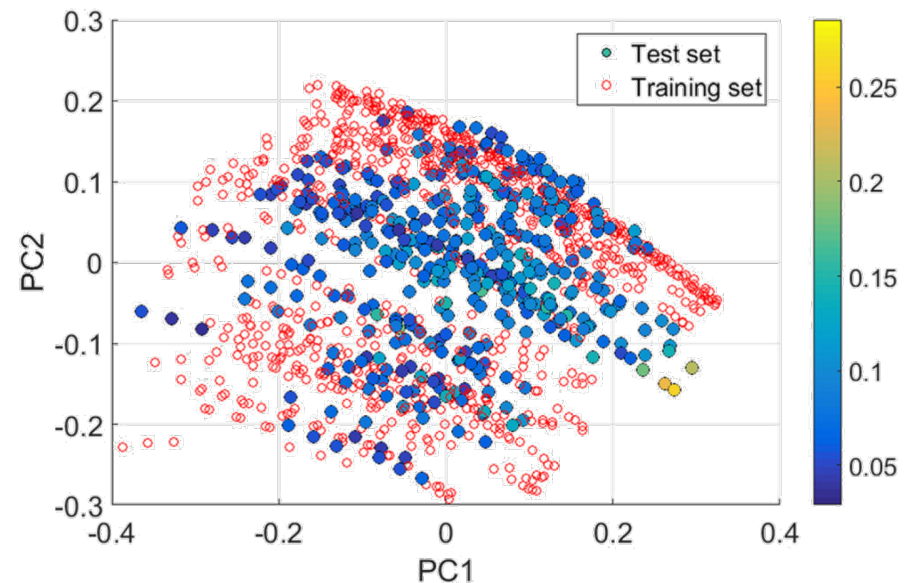
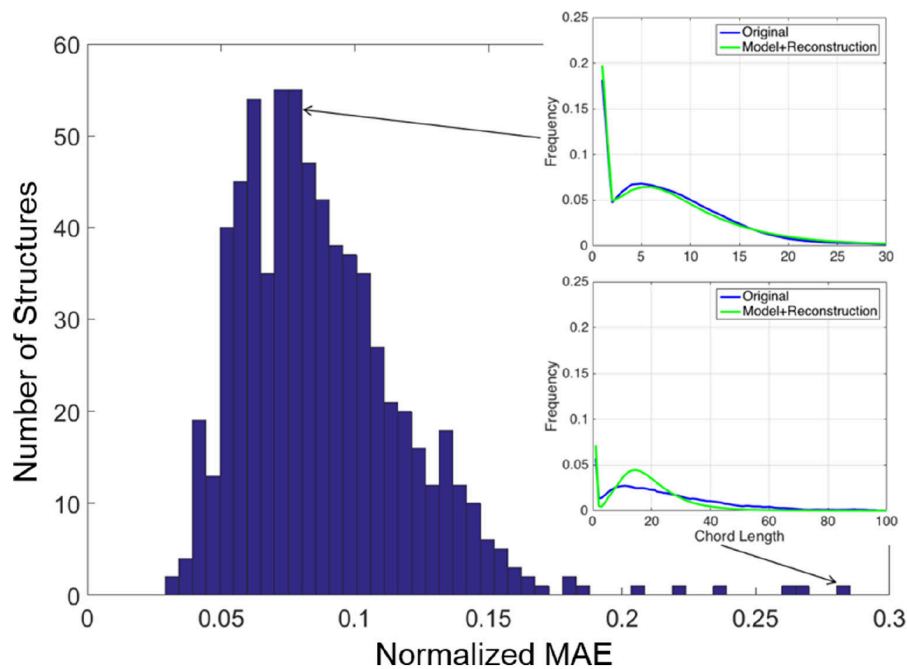
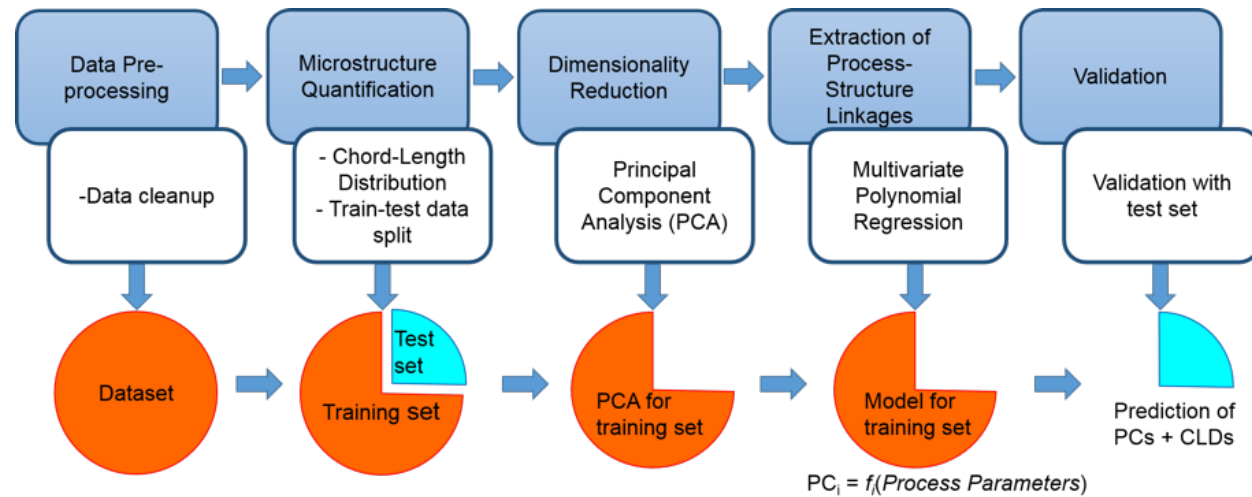


a)



# Exploring AM via Data Science II

- Developed a reduced-order model for microstructure prediction using chord length distributions (CLDs).

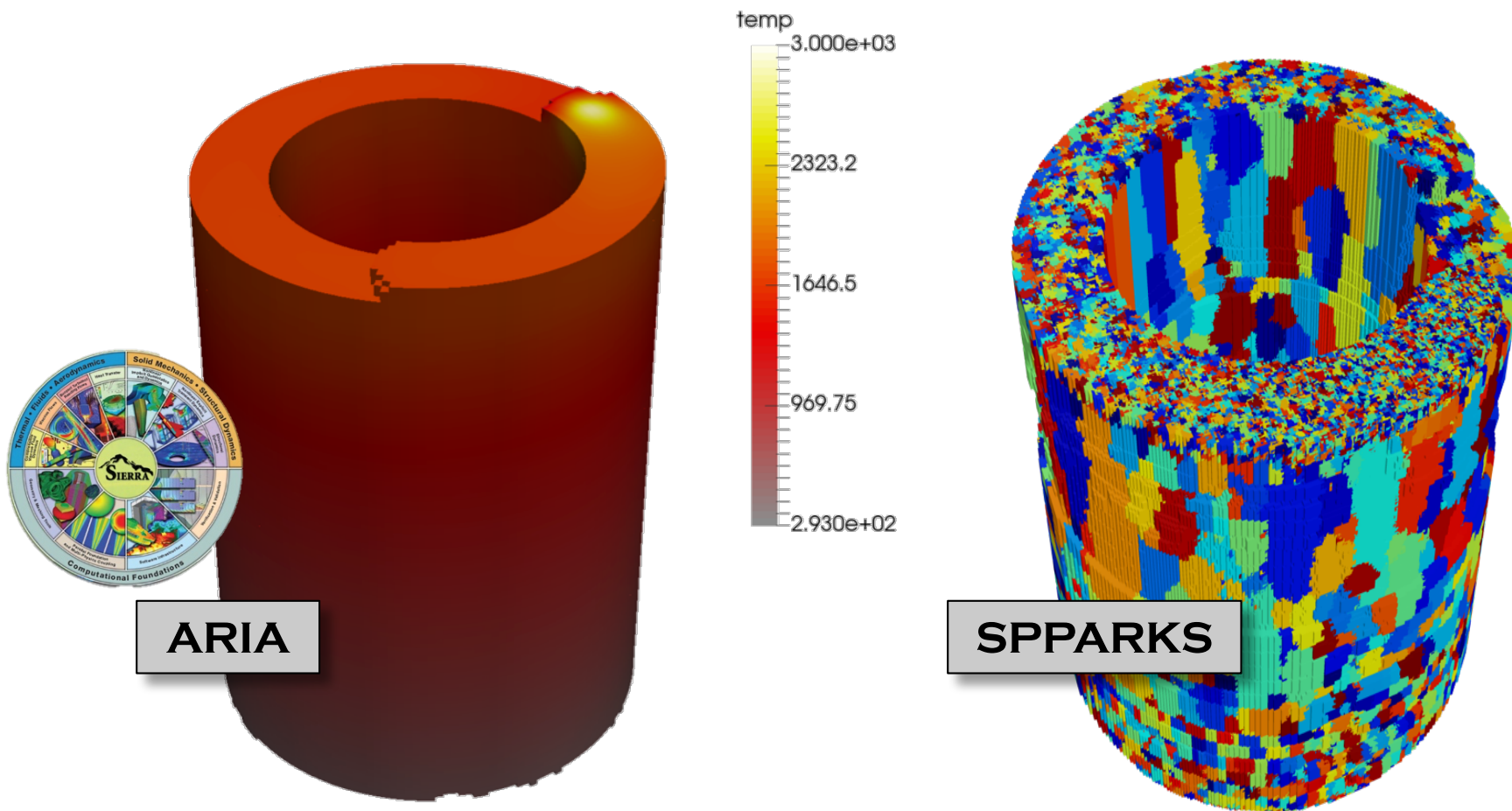


# **Application to Thermal History Models**



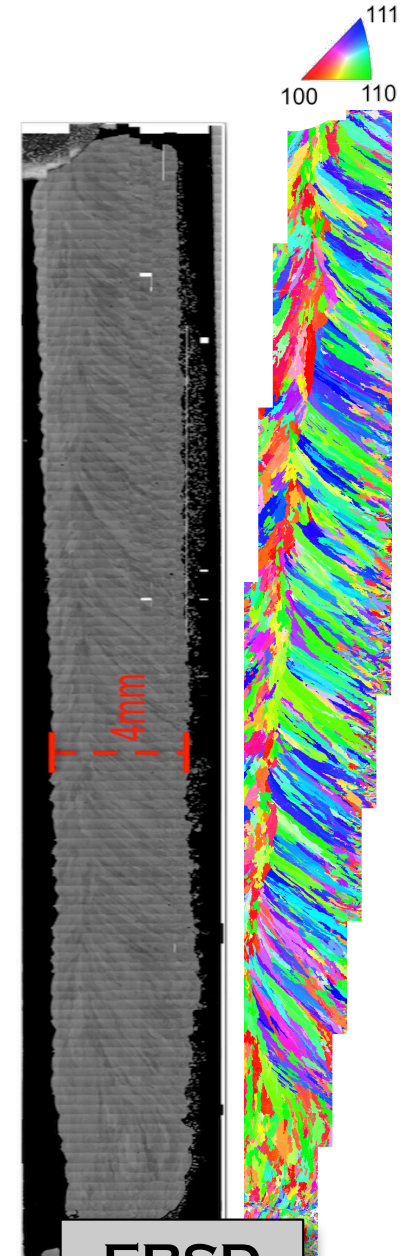
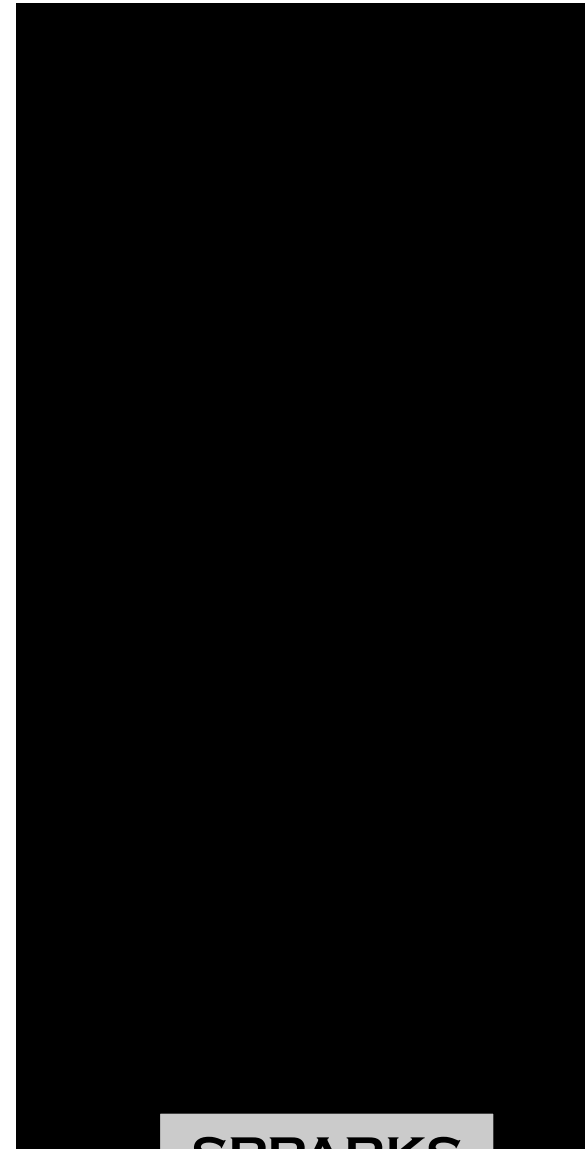
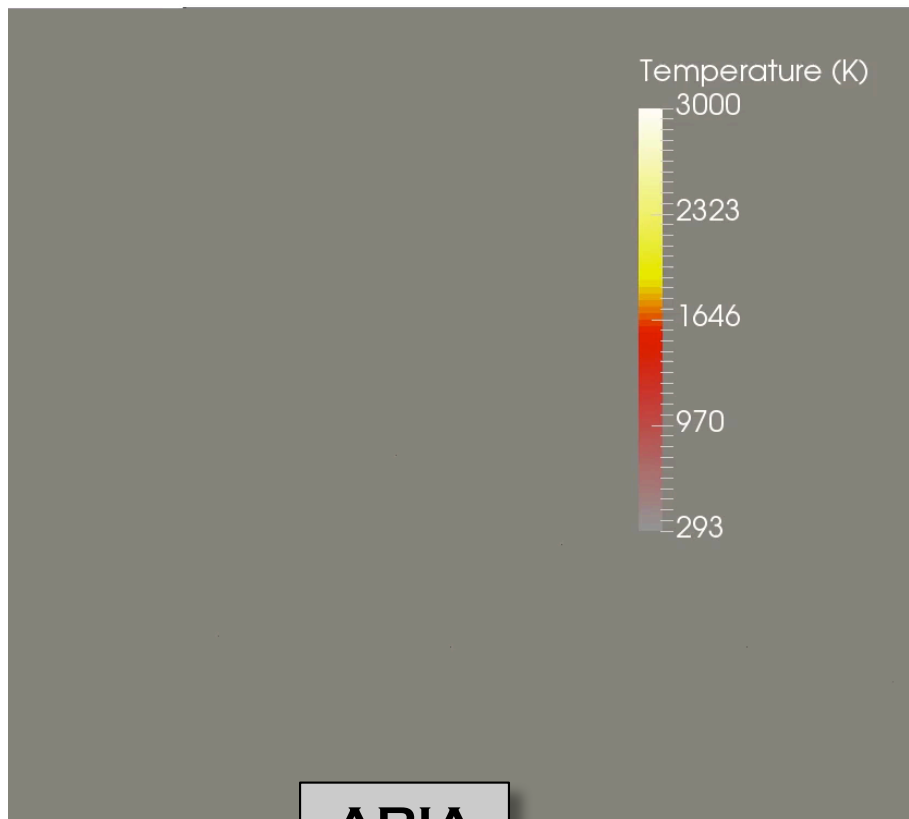
# Linking to Thermal Models I

- Using thermal histories simulated within SNL's production code, ARIA as an input, the modified Monte Carlo Potts model within SPPARKS is coupled to predict grain morphologies within an additive manufacturing build. This demonstrates an ability to predict grain morphology from a rigorous multi-physics model as opposed to a simplified and idealized approximation of the molten zone.



# Linking to Thermal Models II

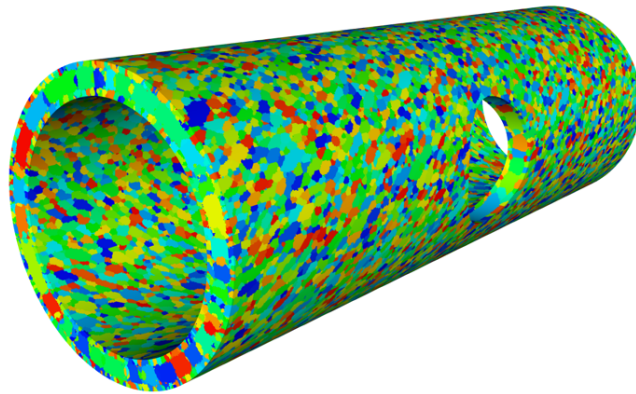
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# **Application to Mechanics**

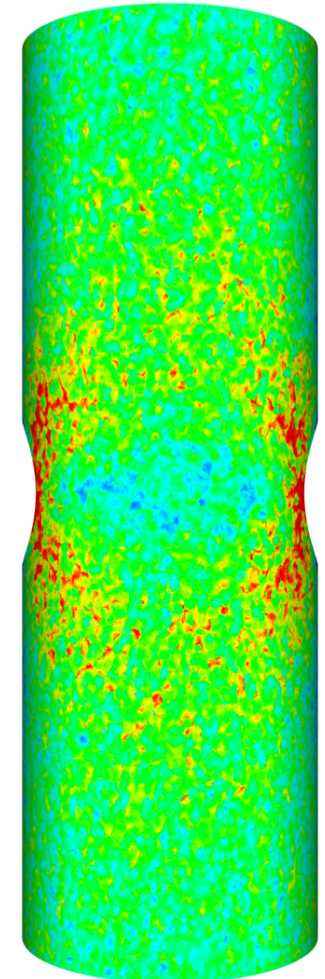
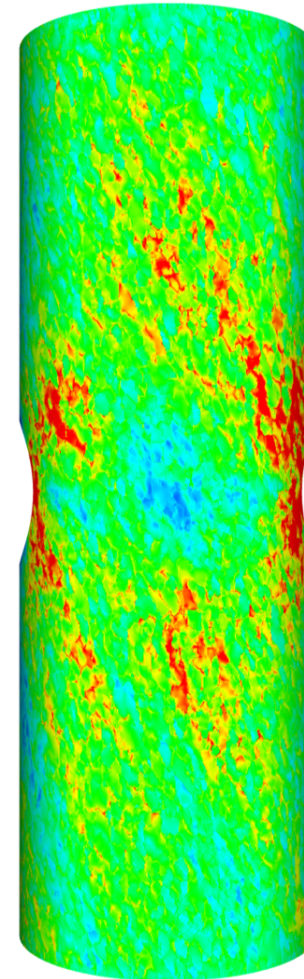
# DNS Inputs for Mechanics

- AM simulations have been performed on cylindrical domains and the resulting microstructures used in solid mechanics simulations.

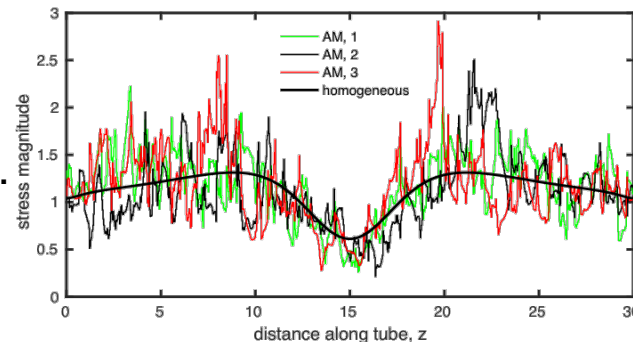
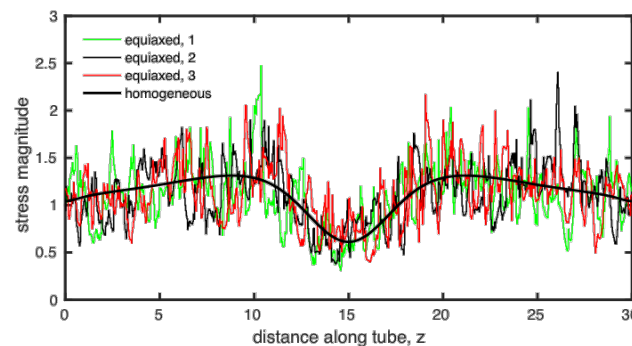


AM-micro stresses

Equiaxed stresses

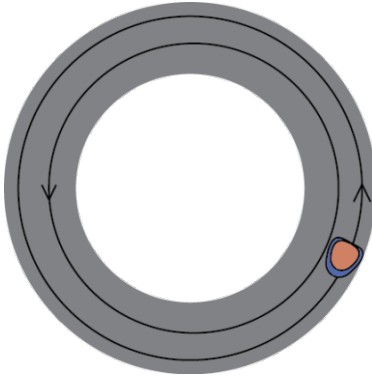


- Stress fields along tube for several equiaxed (top) and AM (bottom) microstructures.



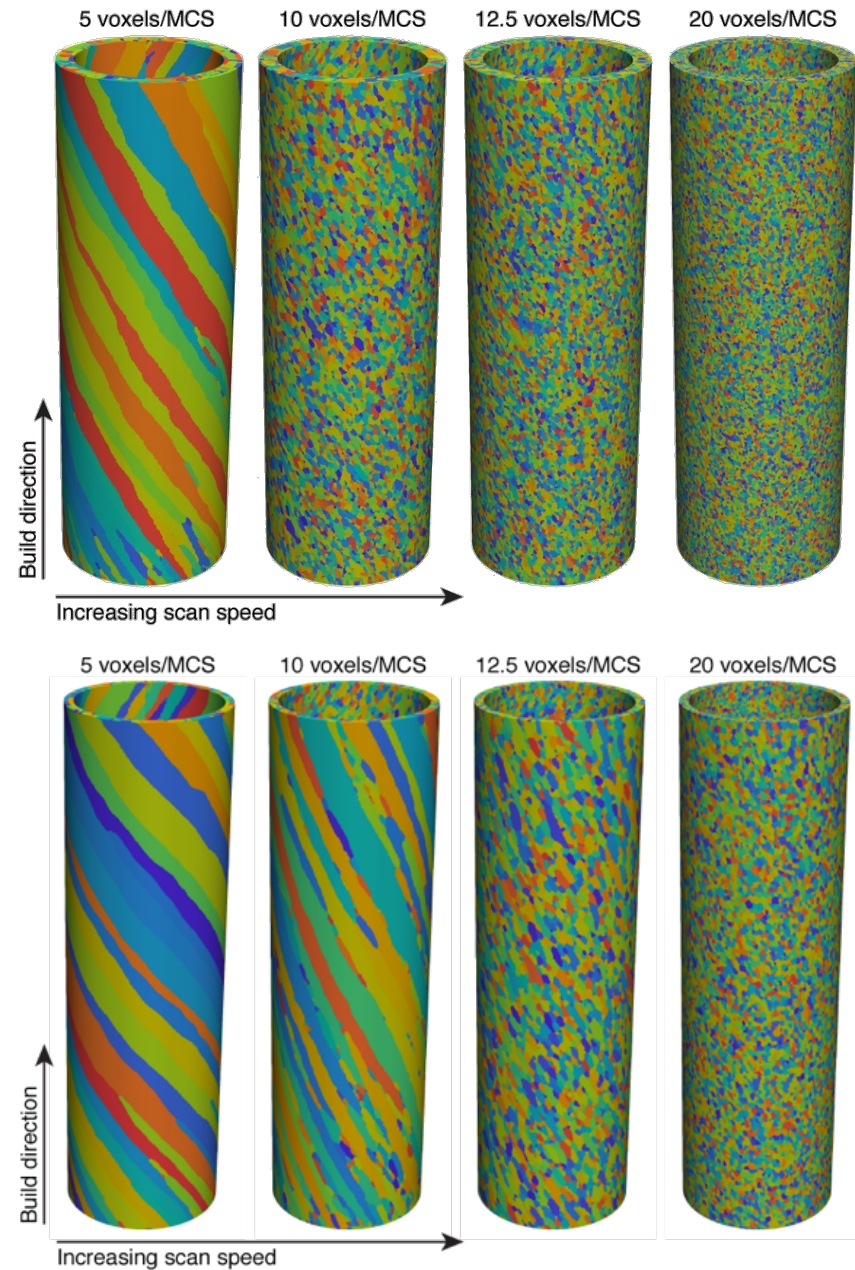


# DNS Inputs for Mechanics



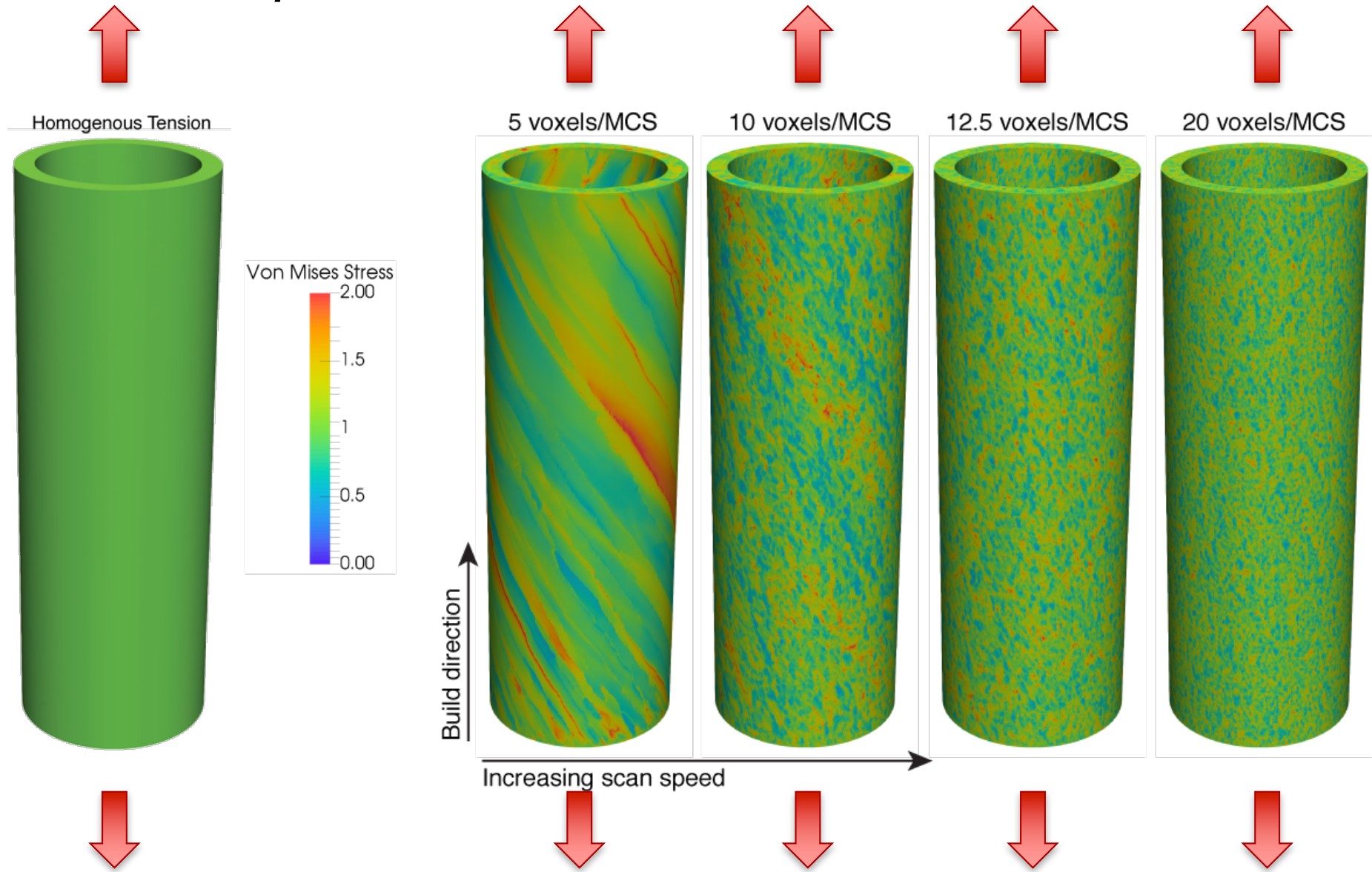
## Synthetic AM builds

- 4 scan velocities.
- 2 concentric circular scan paths per layer.
- Idealized molten pool
- Significant microstructure variation w.r.t. scan velocities and w.r.t. wall thickness.



# Direct Numerical Simulation

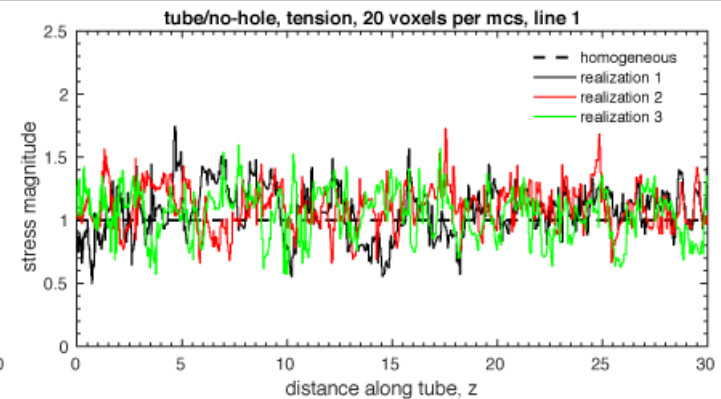
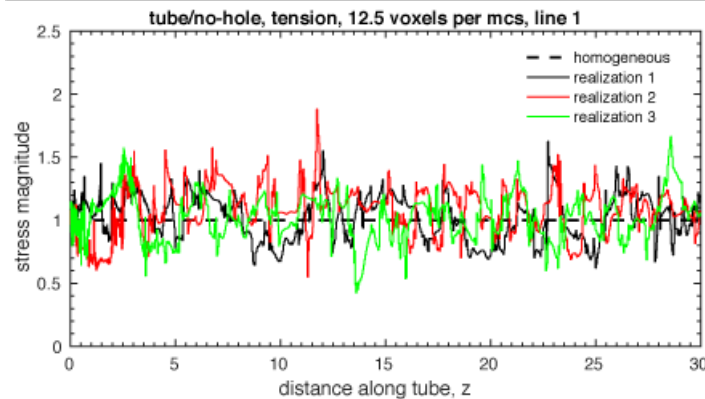
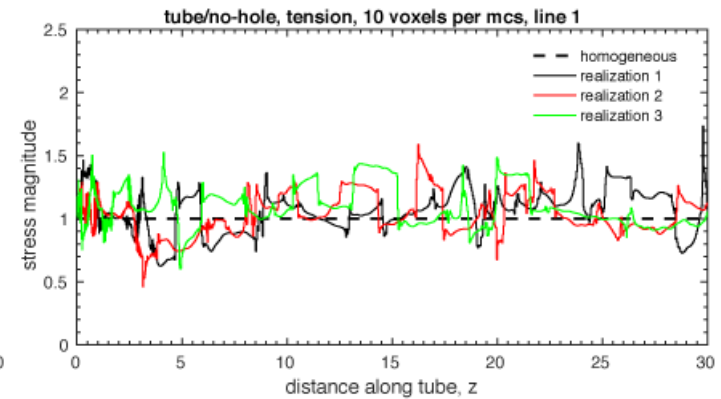
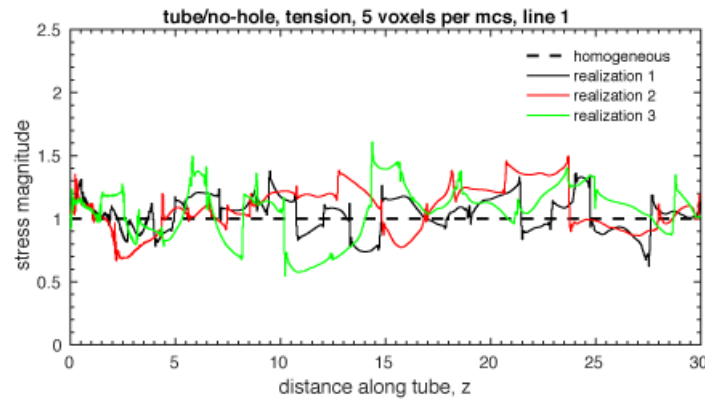
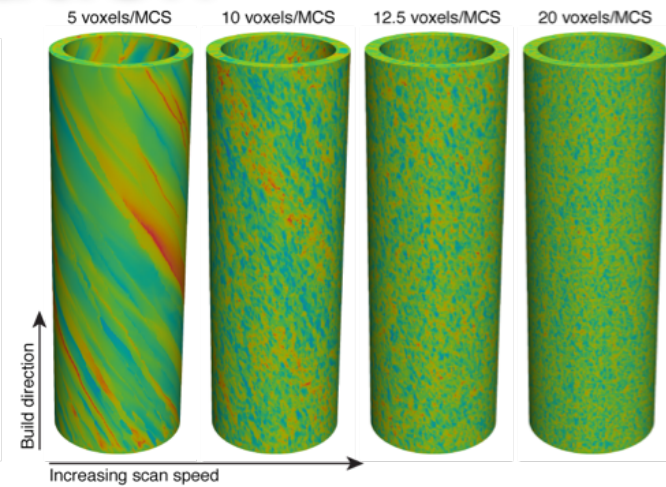
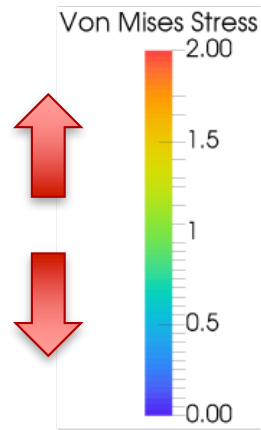
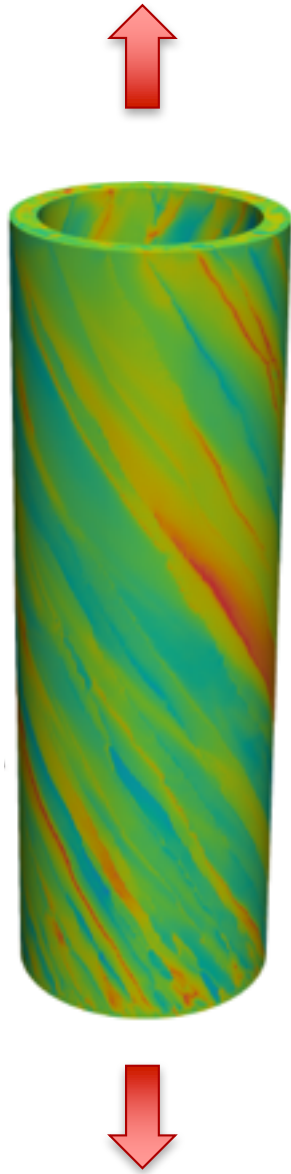
## *Stress response in tension*





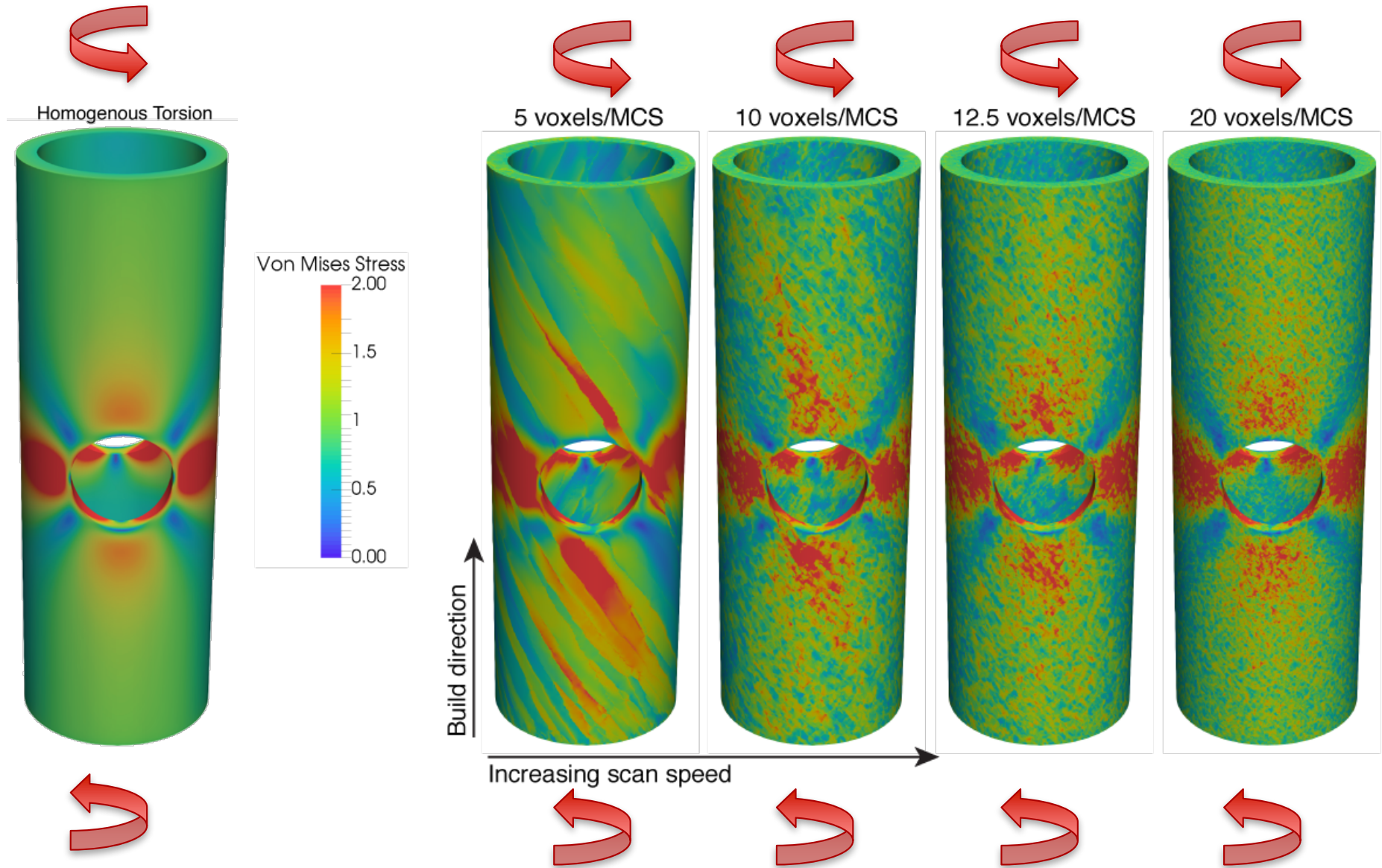
# Direct Numerical Simulation

## Stress response in tension



# Direct Numerical Simulation

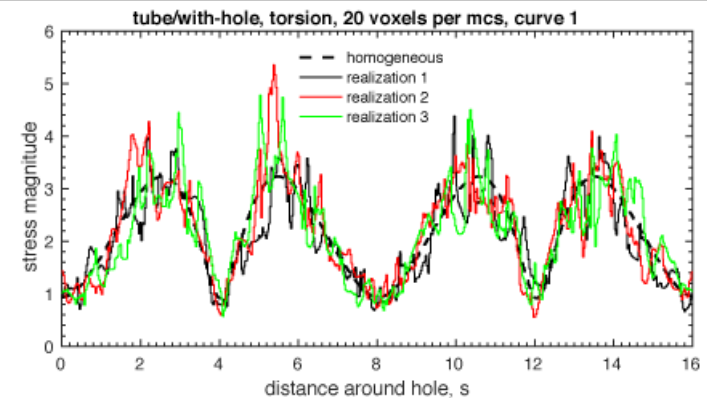
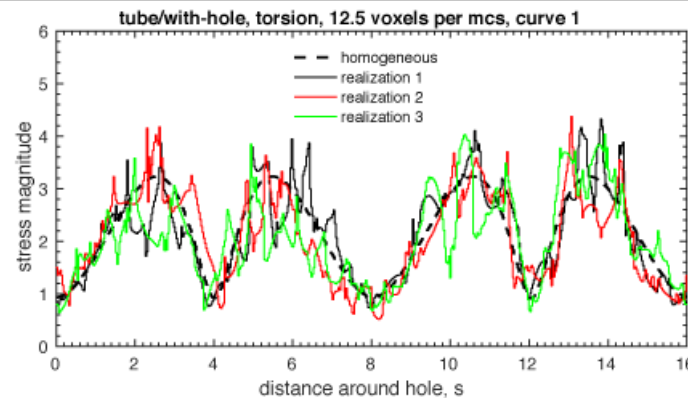
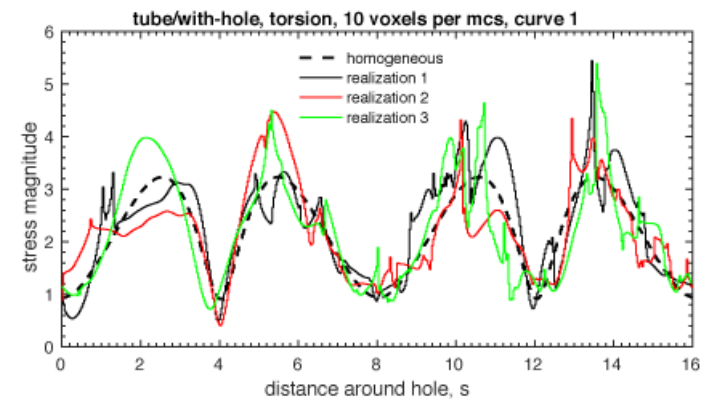
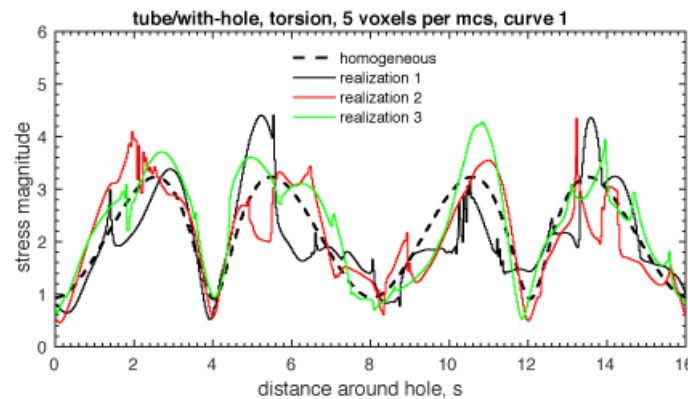
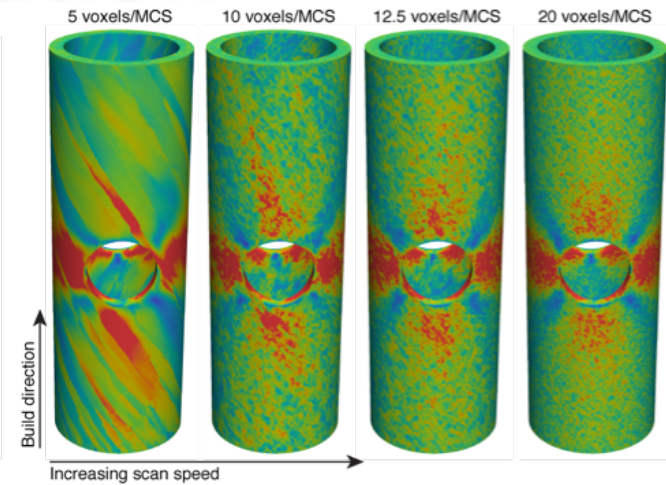
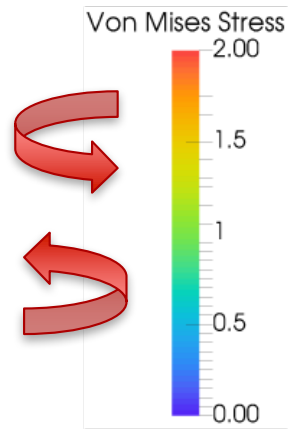
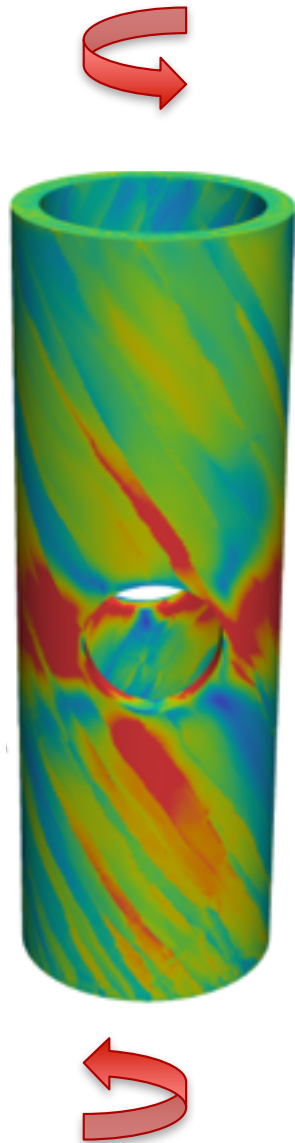
*Stress response in torsion (w/stress concentrator)*



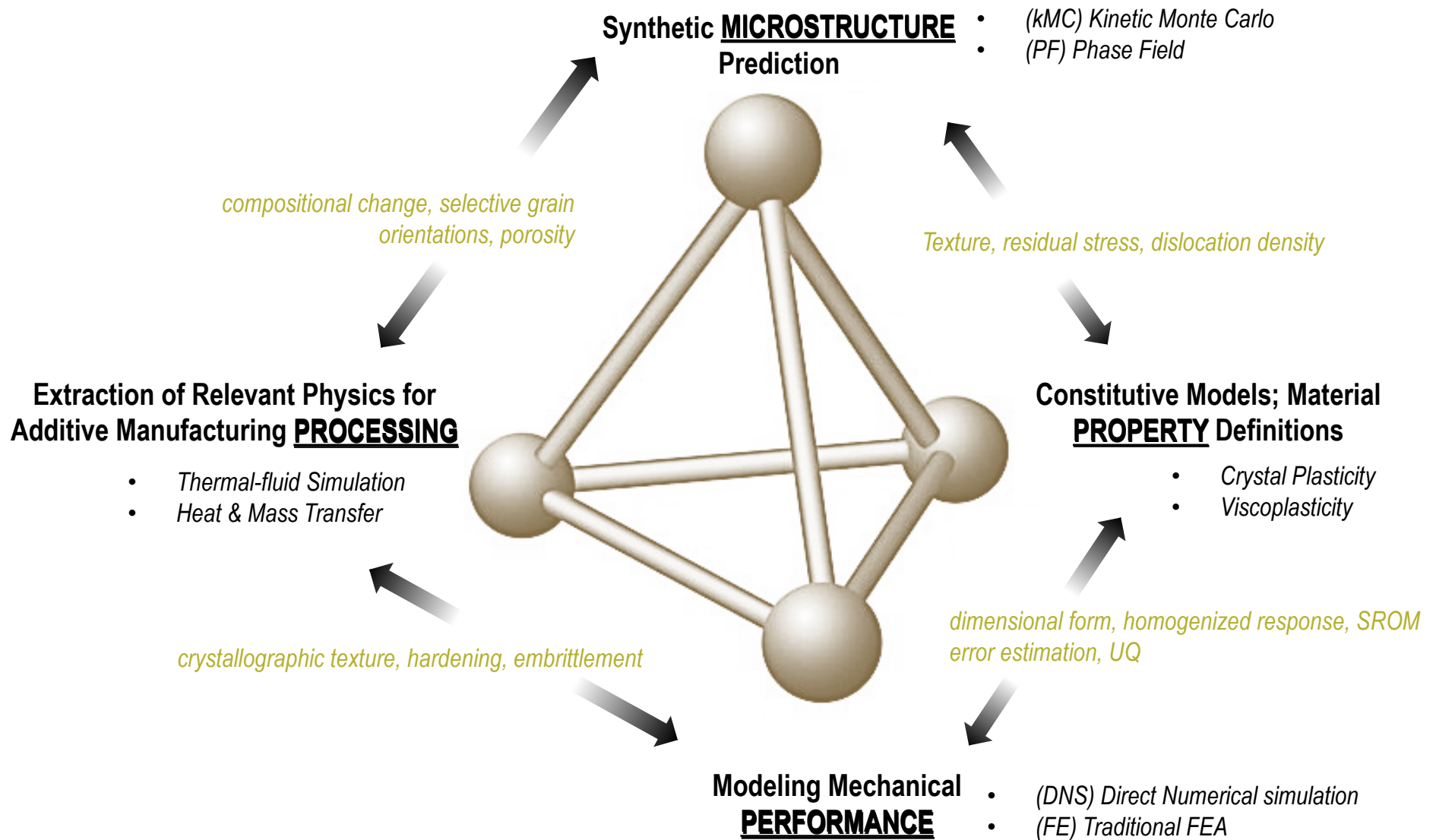


# Direct Numerical Simulation

## Stress response in torsion



# P-S-P-P Linkages by Simulation



# Summary

- **Process-Parameter space in additive manufacturing is so large and so rapidly evolving, traditional design of experiment approaches are inadequate to reliably ascertain conclusive and exhaustive understanding of AM product**
- **Versatile modeling tools are needed to help scope, inform and aid our understanding of AM product**
- **SPPARKS is a versatile, open source, on-lattice Monte Carlo simulation suite containing a variety of microstructural evolution apps and has been adapted to simulate first order approximations of AM microstructure in 2- and 3-dimensions**
- **Synthetic SPPARKS microstructures are now being utilized to:**
  - **Rapidly explore process-parameter space via data science methods**
  - **Link with more rigorous thermal-fluid models**
  - **Serve as inputs for mechanics simulations**

# Questions

1. T. Rodgers, J. Madison, V. Tikare, “*Simulation of Metal Additive Manufacturing Microstructures Using Kinetic Monte Carlo*” **COMPUTATIONAL MATERIALS SCIENCE**, vol. 135, (2017) pp. 78-89
2. E. Popova, T. Rodgers, X. Gong, A. Cecen, J. Madison, S. Kalidindi, “*Process-Structure Linkages Using a Data Science Approach: Application to Simulated Additive Manufacturing Data*” **INTEGRATING MATERIALS AND MANUFACTURING INNOVATION**, vol. 6, (2017) pp. 54-68
3. T. Rodgers, J. Madison, V. Tikare, “*Predicting Mesoscale Microstructural Evolution in Electron Beam Welding*” **JOM**, vol. 68, no. 5, (2016) pp. 1419 – 1426
4. M. Francois, A. Sun, W.E. King, et al., “*Modeling of Additive Manufacturing Processes for Metals: Challenges and Opportunities*” **CURRENT OPINION IN SOLID STATE AND MATERIALS SCIENCE**, (2017) *in press*
5. J. Madison, “*Integrated Computational Materials Engineering: Tools, Simulations and New Applications*” **JOM**, vol. 68, no. 5, (2016) pp. 1376-1377
6. <http://spparks.sandia.gov>

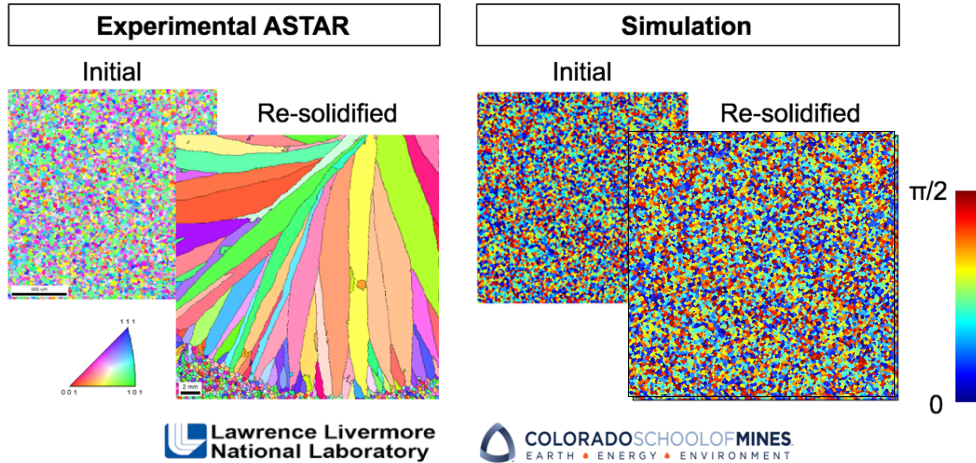




# Backup Slides

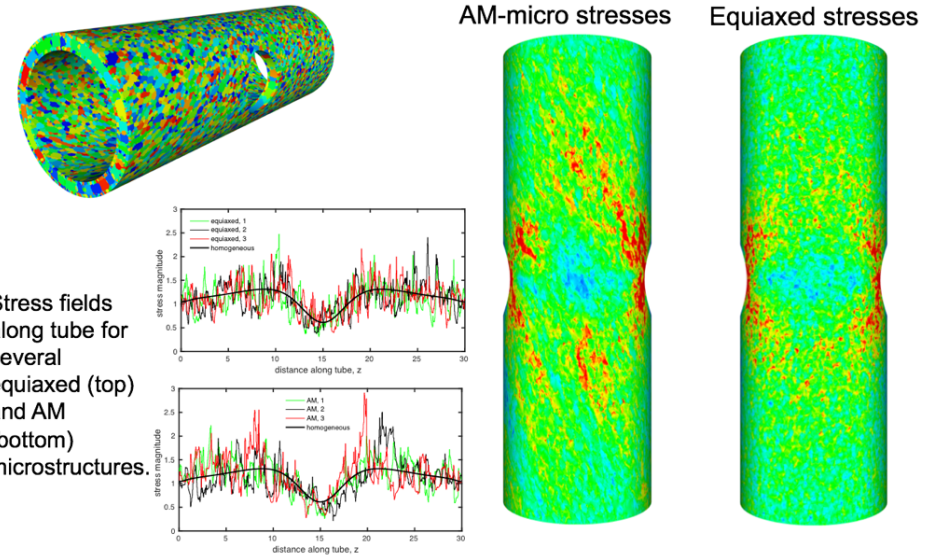
# Grain Growth in Thin Films

- Thin Films of Al-Si binary alloys melted to undergo rapid solidification while being observed in a time-resolved dynamic TEM (DTEM).
- Solidification interface profiles are extracted and used as inputs for a 2D kMC model.
- Simulations are now being modified to allow incorporation of crystallographic texture



# Inputs for DNS for Mechanics

- AM simulations have been performed on cylindrical domains and the resulting microstructures used in solid mechanics simulations.

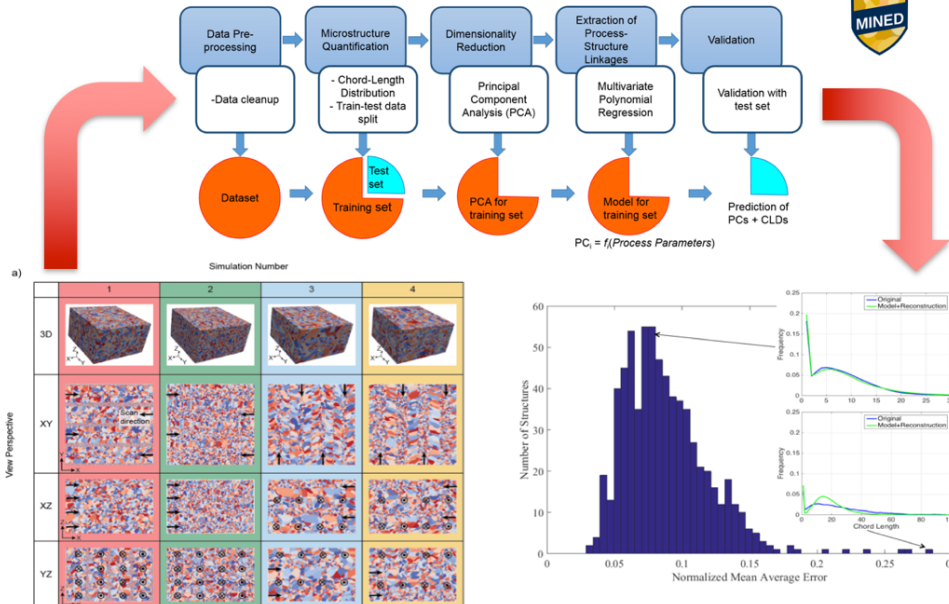


J. McKeown (LLNL), A. Clarke (CSM), J. Gibbs (LANL), T. Rodgers, J. Madison (SNL)

J. Bishop, T. Rodgers, J. Madison, (Sandia National Labs)

# Exploring AM via Data Science

- Developed a reduced-order model for microstructure prediction using chord length distributions (CLDs).



# Linking to Thermal Models

- Using thermal histories simulated within SNL's production code, ARIA as an input, the modified Monte Carlo Potts model within SPPARKS is coupled to predict grain morphologies within an additive manufacturing build. This demonstrates an ability to predict grain morphology from a rigorous multi-physics model as opposed to a simplified and idealized approximation of the molten zone.

