

***Project 36B-L: Rationalization of Liquid/Solid
and Solid/Solid Interface Instabilities During Thermal-
Mechanical Transients of Metal Additive Manufacturing
(ISU)***

***Semi-annual Fall Meeting
October 2021***

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- Faculty: Dr. Peter Collins (ISU)
- MURI members: S. Babu (UTK), A. Clarke (CSM), J. R. Jinschek (OSU), T. Pollock (UCSB), Z. Kong (VT), S. Ringer and X. Liao (Univ. of Sydney), S. Primig (Univ. of South Wales) and TEAM
- Industrial Mentors: Billy Short and Jennifer Wolk (ONR)

Project 36B-L: Rationalization of Liquid/Solid and Solid/Solid Interface Instabilities During Thermal-Mechanical Transients of Metal Additive Manufacturing (ISU)



<ul style="list-style-type: none"> ISU team: Katie O'Donnell, Maria Quintana, Amamchukwu Ilogebe, Matt Kenney Advisor(s): Prof. Peter Collins (ISU) 	<p>Project Duration August 2018 to August 2021 Extended: August 2021 to August 2023</p>
<ul style="list-style-type: none"> Problem: Understand the thermal gradients in an AM build as a function of different scan strategies by studying the microstructure. Objective: To understand the science behind the relation between thermal gradients and the microstructure and texture evolution. Benefit: Optimize the final cost and mechanical properties of the AM component. 	<p>Recent Progress</p> <ul style="list-style-type: none"> Statistical analysis of finite element simulations of stress distribution on gas pores Parent β grain reconstructions Analysis of local property variations (micro- and nanoindentation) Preliminary microstructural analysis of Haynes 282 samples

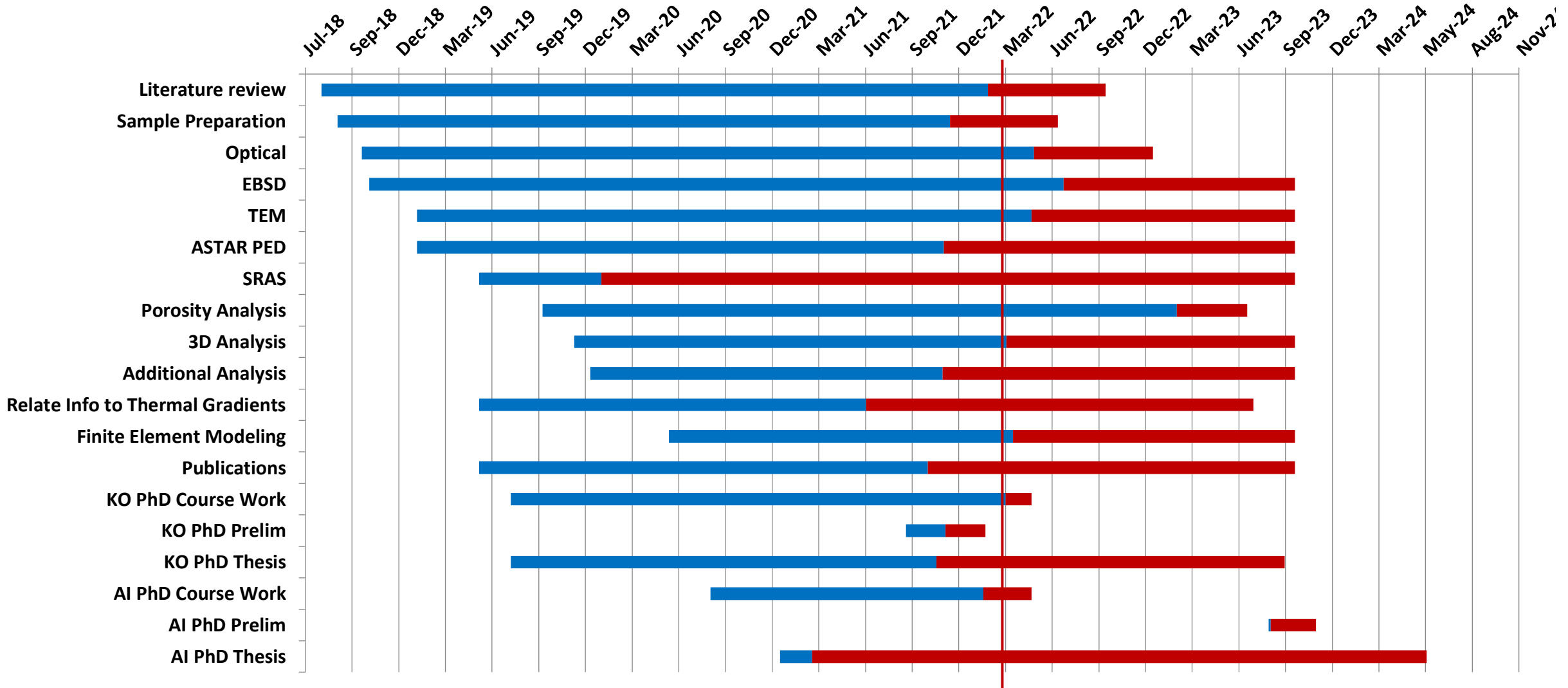
Metrics		
Description	% Complete	Status
1. Sample preparation for optical, SEM-BSE, EBSD and TEM	75%	●
2. Literature review	80%	●
3. Texture scans – EBSD, SRAS, and ASTAR PED	65%	●
4. 3D analysis	20%	●
5. Relate thermal gradients to microstructure and the final mechanical properties	55%	●

Industrial Relevance



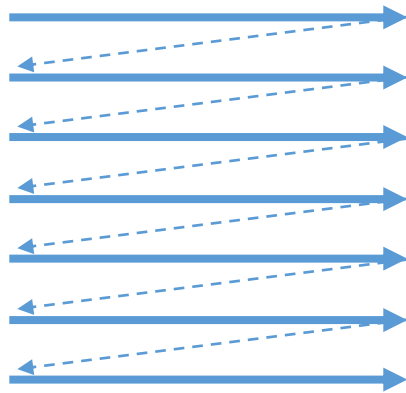
- Understanding underlying behavior of different AM strategies on resulting microstructure and mechanical properties of metallic printed parts
- Build a scientific basis into Integrated Computational Materials Engineering (ICME) predictions of AM knowledge gap areas (nano and micro scale regimes of length and time)
- Reduce trial and error phase of AM design and manufacture curve

Progress



Research Interests

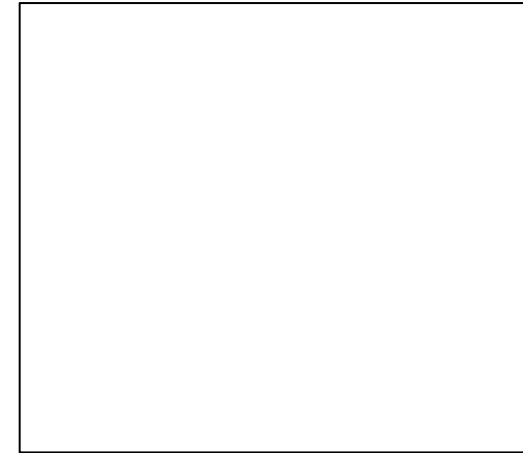
Three different AM scan strategies are selected to understand fundamental research questions. The different scan strategies will change the thermal gradient: Raster, Dehoff, and Random



Ordering of
Raster Fill



Ordering of
Dehoff Fill



Ordering of
Random Fill

Outline of the Project



Ti64, Inconel 738, and Haynes 282 builds with different scan strategies and different geometries are provided by ORNL – Raster (L), Random (R), and Dehoff (D)

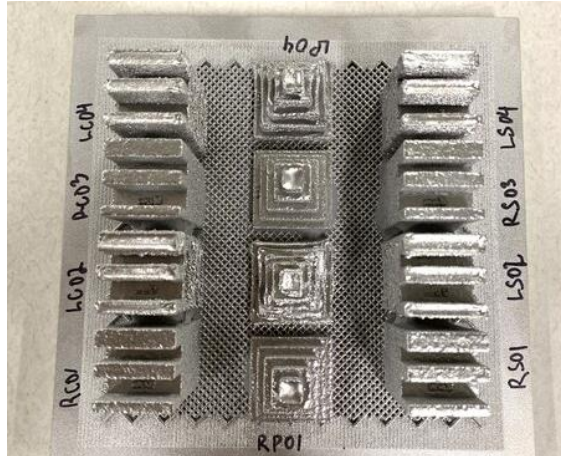
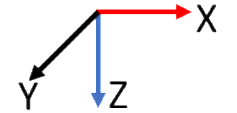
TASKS:

Imaging – Macro, Optical, and SEM-BSE

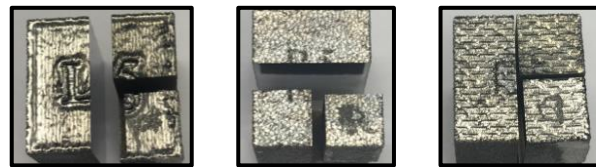
1. Texture (across length scales)
 - A. SRAS – Spatially Resolved Acoustic Spectroscopy (macro-scale)
 - B. EBSD – Electron Back Scattered Diffraction (SEM) (micro-scale)
 - C. PED – Precession Electron Diffraction (TEM) (nano-scale)
2. Analysis of the 2D and 3D data
3. Develop the understanding to relate thermal gradient to the microstructural evolution

Material

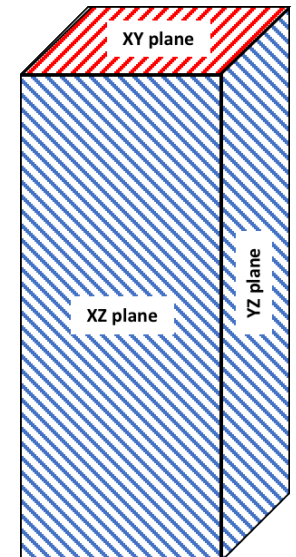
- 3 cuboid Ti64 builds – Raster (L5), Random (R5), and Dehoff (D5)
- 2 cuboid Inconel 738 builds – Random and Raster
- 3 Haynes 282 samples – pyramid, printed "pores" (cube, spiral)
- Z is the build direction for all the samples



Haynes



Raster (L5) Random (R5) Dehoff (D5)
Ti64



15x15x25mm

Outline



- Statistical analysis of results from finite element modeling
- Parent β grain reconstruction of EBSD maps from Ti-6Al-4V
- Local property variations from Ti-6Al-4V (micro- and nanoindentations)
- Initial microstructure results from Haynes samples

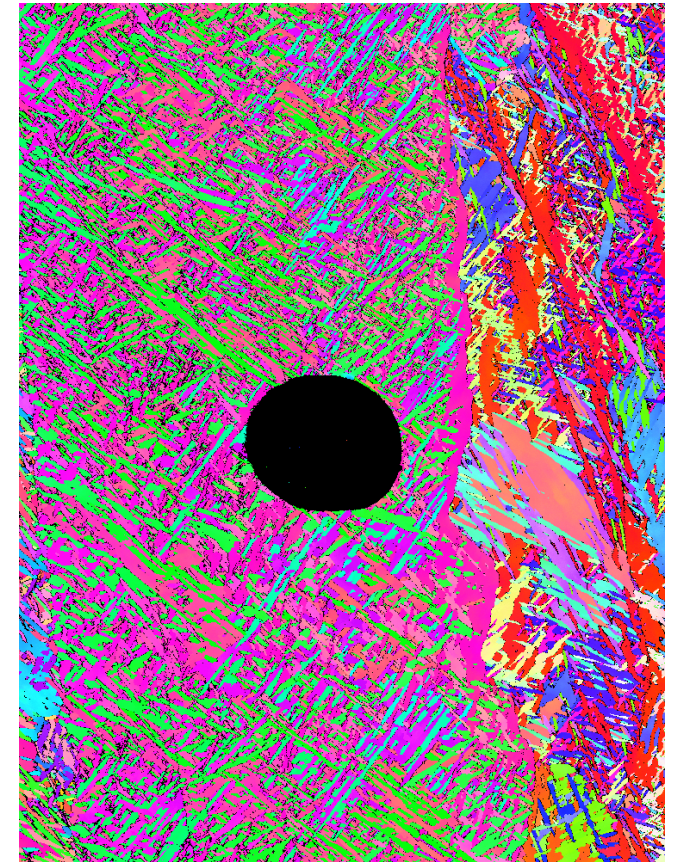
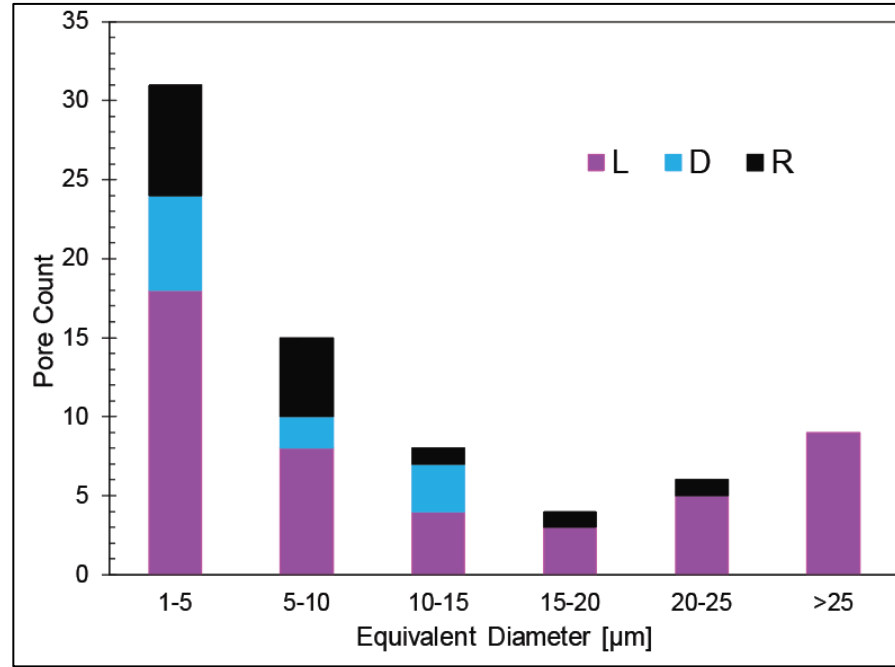
Previously:

Spherical pores ranged from 1-50 μm in diameter

Aluminum variations were present around pores

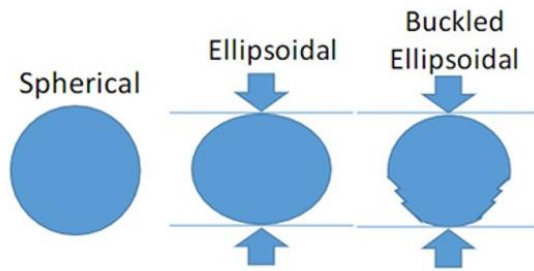
Compression and buckling occurred, likely due to thermal stresses before the final $\beta \rightarrow \beta + \alpha$ phase transformation

Pore size distribution across scan strategies

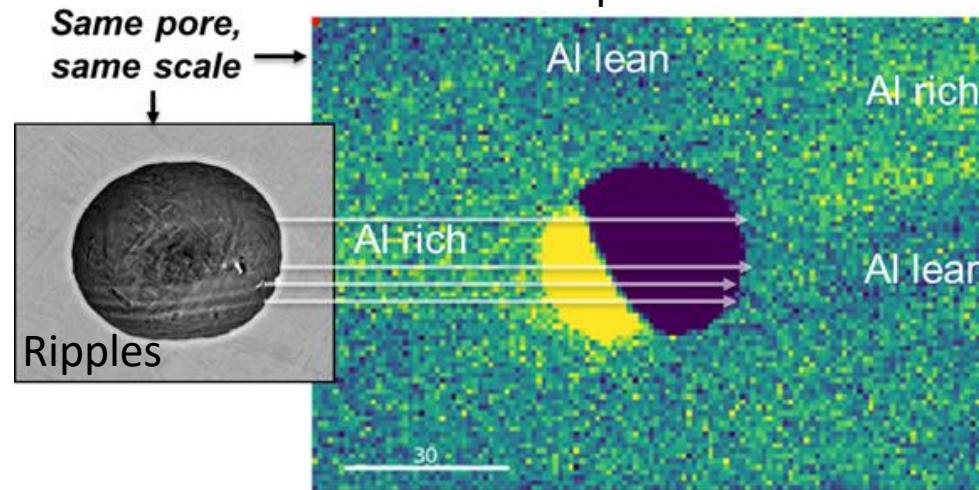


No evidence of deformation in microstructure

Compression and buckling



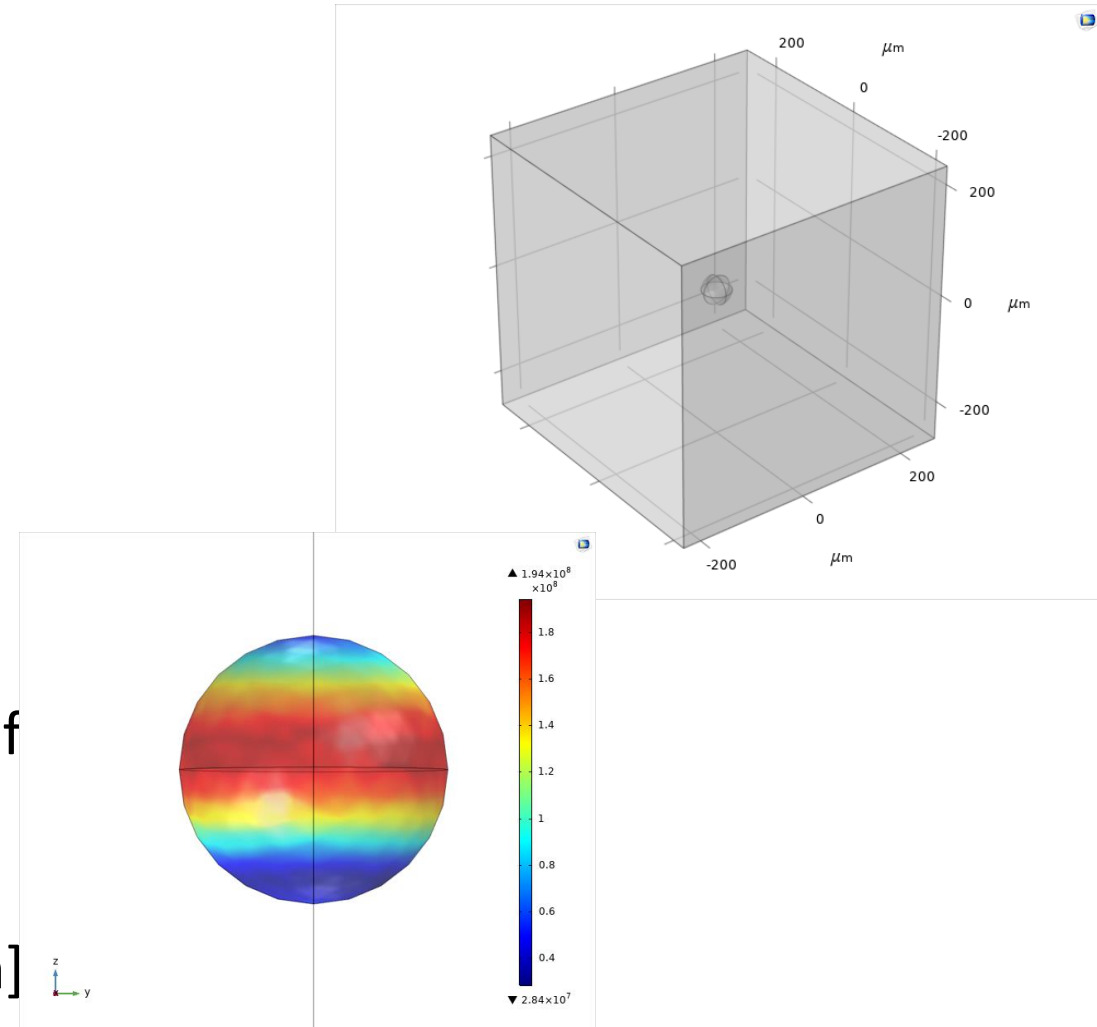
Compositional Variations



Finite Element Modeling Thermal Gradient DOE

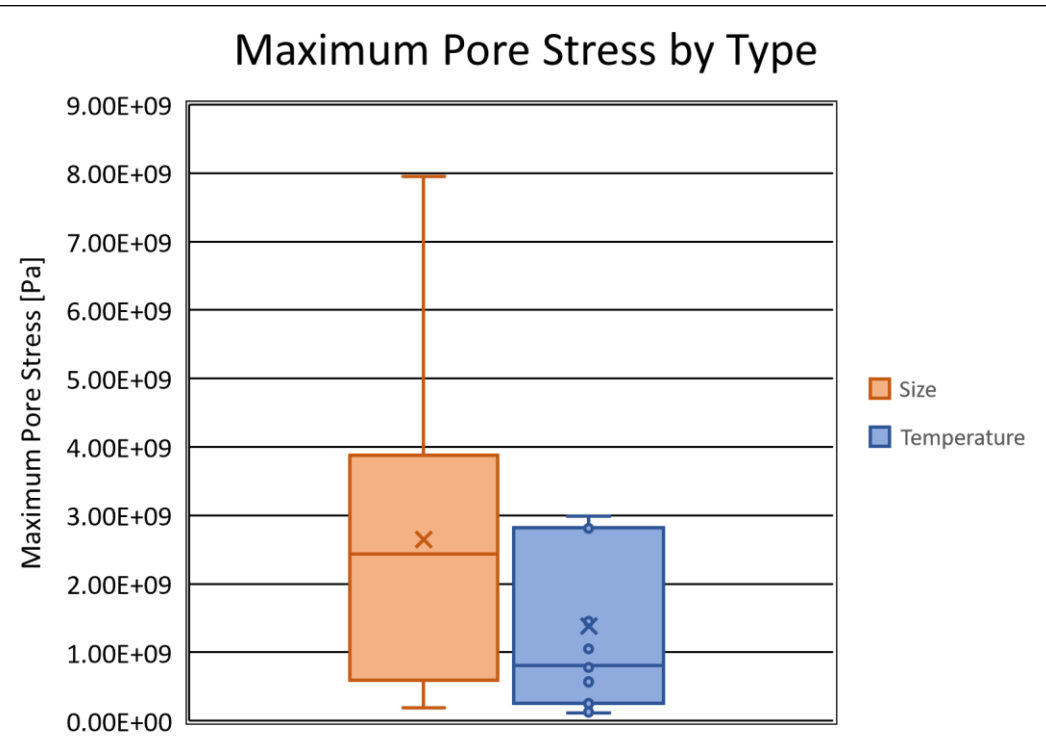
In-depth understanding of thermal gradients and pores

- Variables:
 1. Constant size matrix [100 μm], varying temperature fields OR constant temperature fields [1600 $^{\circ}\text{C}$ top, 470 $^{\circ}\text{C}$ bottom], varying matrix size
 2. Pore location within the matrix [1/3, 1/2, or 2/3 height] (temperature field of the pore)
 3. Pore size [5, 10, 25, or 50 μm diameter]
 4. Thermal gradient [0.5, 1, 5, or 10 $^{\circ}\text{C}/\mu\text{m}$]

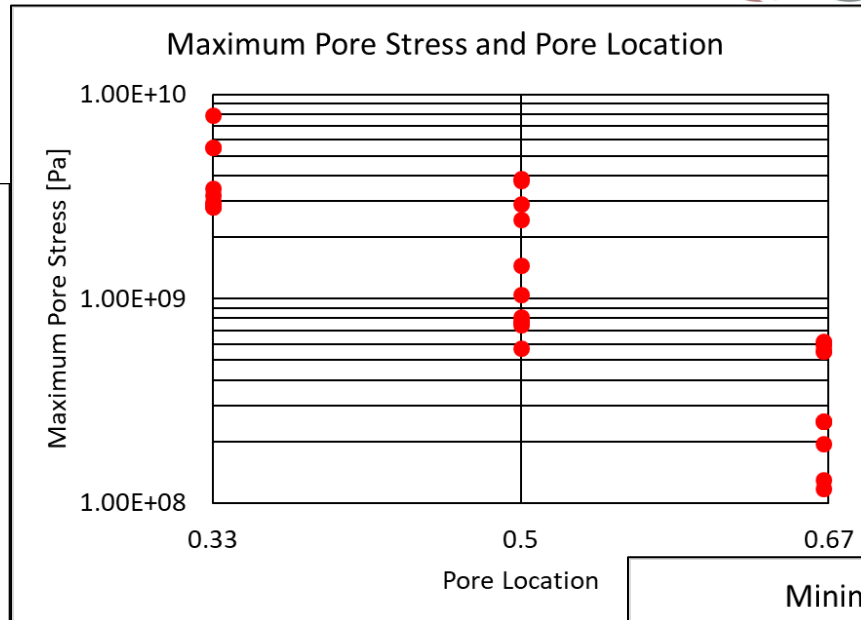


Finite Element Modeling Thermal Gradient DOE

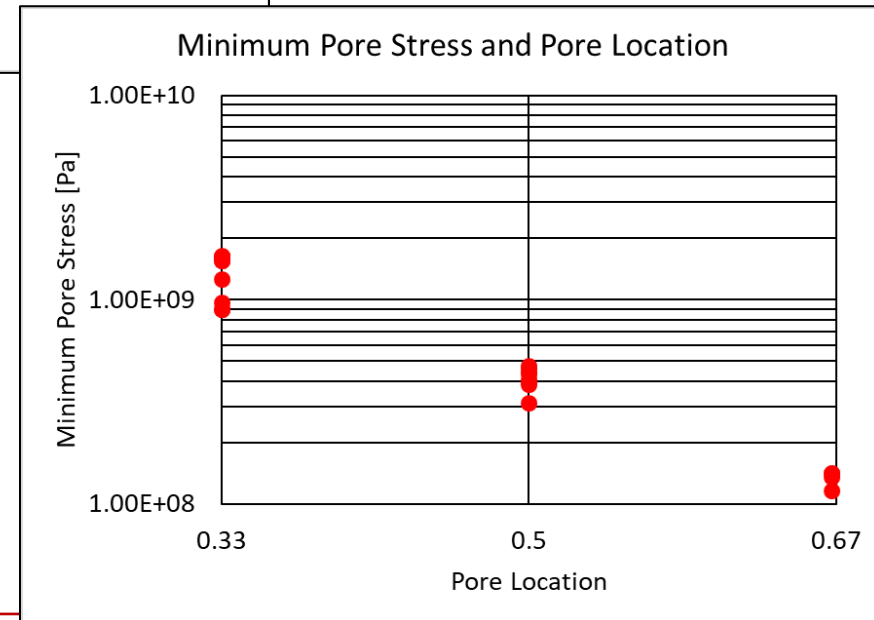
Looking only at simple data analysis...



...we can see that smaller matrix sizes (i.e., constant matrix sizes) lead to higher stresses...



...and that pore location (i.e., the temperature field of the pore) seems to have the largest effect on resulting pore stress.



Finite Element Modeling Thermal Gradient DOE

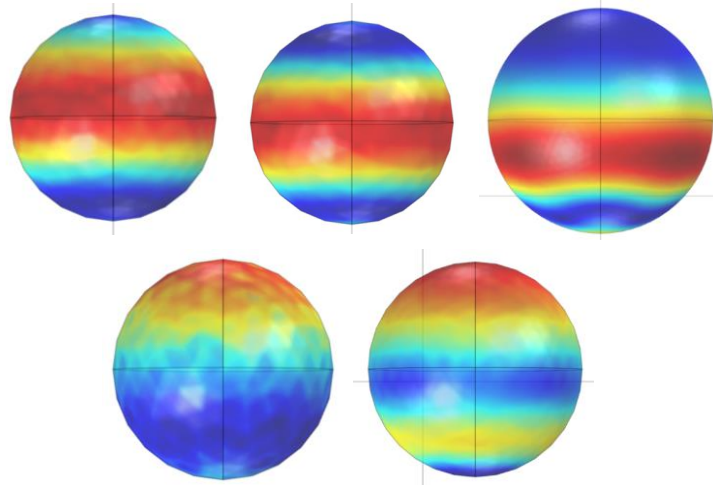
Stress distribution type also seems to be most affected by pore location (i.e., the temperature field of the pore)

- There is no overlap in types found in pores at 2/3 of the height of the matrix as compared to types found in pores at the other two locations.

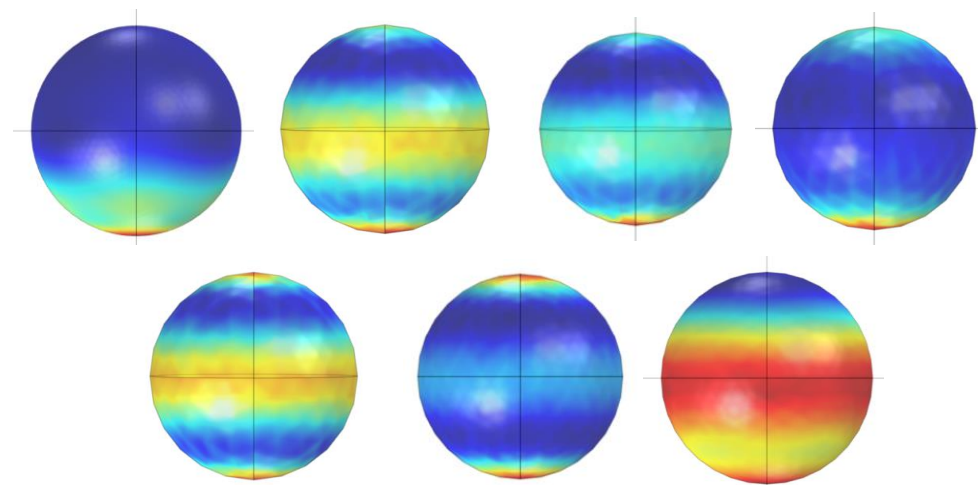
Pore locations:

- 2/3
- 1/2
- 1/3

Max stress at middle or top only
Pore found only at 2/3 height



Max stress at bottom, bottom and top,
or bottom and middle only
Pores found only at 1/3 or 1/2 height



Finite Element Modeling Thermal Gradient DOE – ANOVA



Analysis of variance (ANOVA) is used “to determine the influence that independent variables have on the dependent variable”¹.

For these results, we can rely on the p-value to determine statistically significant relationships, where results where $p < 0.05$ are statistically significant.

Individual Analysis

Max Pore Stress against each variable individually

Variable	p-value
Constant Factor	0.122
Pore Size	0.123
Thermal Gradient	0.263
Pore Location	7.85E-8

Significantly lower than all other values

R Studio code: `X <- aov(Max.Pore.Stress ~ variable, data = filename)`

We can also analyze additive or interaction effects (Pore Location + Pore Size/Pore Location * Pore Size).

Performing all possible combination of these in relation to Max Pore Stress gives us that the interactions of Pore Location, Pore Size, and Constant Factor produces the best fit.

(Max Pore Stress ~ Pore Location*Pore Size*Constant Factor)

Finite Element Modeling Thermal Gradient DOE



Tukey's Honestly Significant Difference (HSD) tests allow us to test for pairwise comparisons as well.

```
$`as.factor(Pore.Location)`
```

	diff	lwr	upr	p adj
0.5-0.33	-2095083333	-2982106415	-1208060252	0.0000101
0.667-0.33	-3361833333	-4248856415	-2474810252	0.0000000
0.667-0.5	-1266750000	-2153773082	-379726918	0.0041362

Pore Location:
Statistically significant
between all locations.

```
$`as.factor(Pore.Size)`
```

	diff	lwr	upr	p adj
10-5	-96186508	-1305026655	1112653639	0.9962333
25-5	631888889	-498877528	1762655305	0.4331316
50-5	1341027778	262884302	2419171253	0.0106981
25-10	728075397	-480764750	1936915544	0.3683650
50-10	1437214286	277449152	2596979420	0.0110297
50-25	709138889	-369004586	1787282364	0.2940586

Pore Size:
Statistically significant only
between sizes 50-10µm and
50-5µm.

```
$`as.factor(Thermal.Gradient)`
```

	diff	lwr	upr	p adj
1-0.5	138357292	-1175474558	1452189142	0.9913969
5-0.5	712157292	-360581922	1784896505	0.2864741
10-0.5	816370739	-231704506	1864445983	0.1681464
5-1	573800000	-740031850	1887631850	0.6334620
10-1	678013447	-615758723	1971785617	0.4882132
10-5	104213447	-943861798	1152288692	0.9927307

Thermal Gradient:
No statistical significance.

```
$`as.factor(Constant.Factor)`
```

	diff	lwr	upr	p adj
Temperature-Size	-1071691705	-1670799758	-472583651	0.0010792

Constant Factor:
Statistical significance.

Finite Element Modeling Thermal Gradient DOE



In summary...

- The resulting maximum von Mises stress on the surface of gas pores depends much less on the overall thermal gradient than it does the temperature field it is surrounded by*
 - Pores at the lowest temperature field experience the highest stress
 - *Keeping in mind these are stationary snapshots at a single thermal gradient, and do not show the evolving thermal cycles involved in real AM processes
- Pore sizes in future modeling can be restricted to a two-factor variable (5 vs. 50 μ m in diameter)

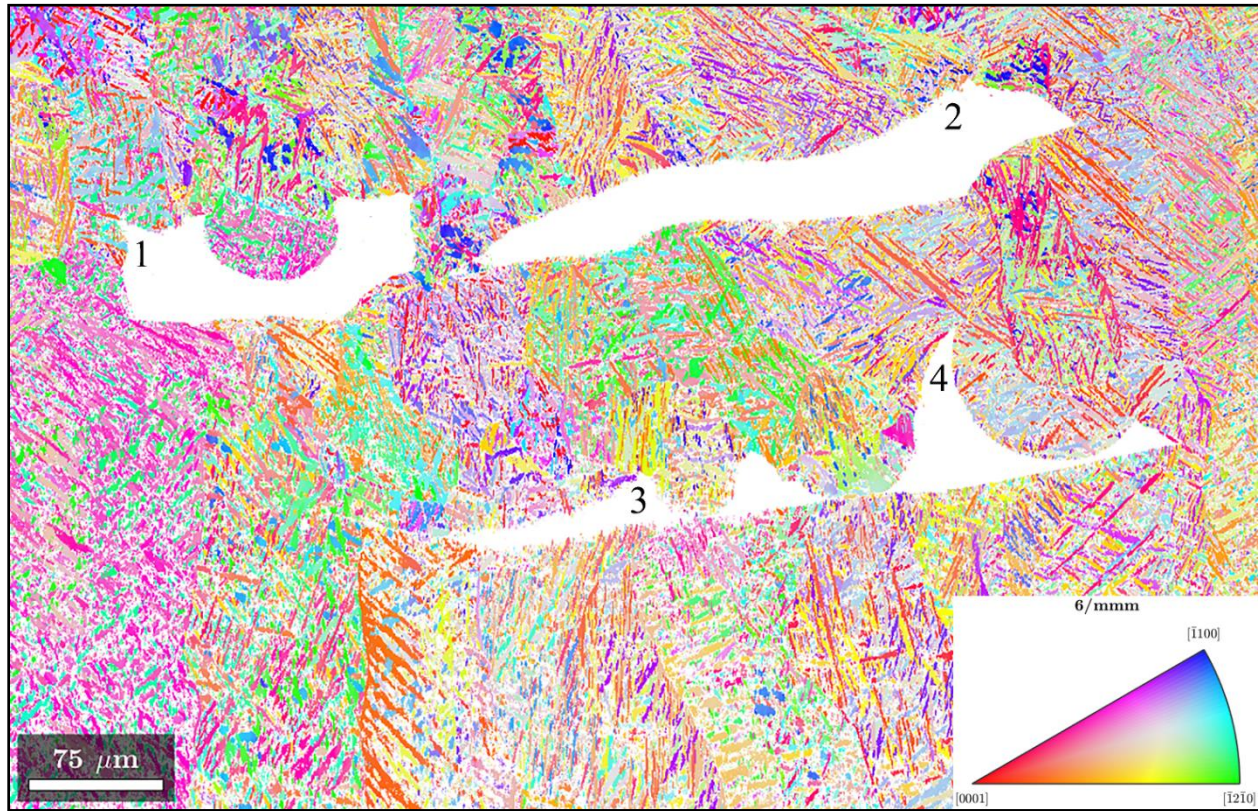
Moving forward...

- Incorporate plasticity into all models
- Evaluate other output parameters beyond von Mises stress
- Expand beyond spherical pores
- Examine the effects of the presence of other pores
- Include chemical variations

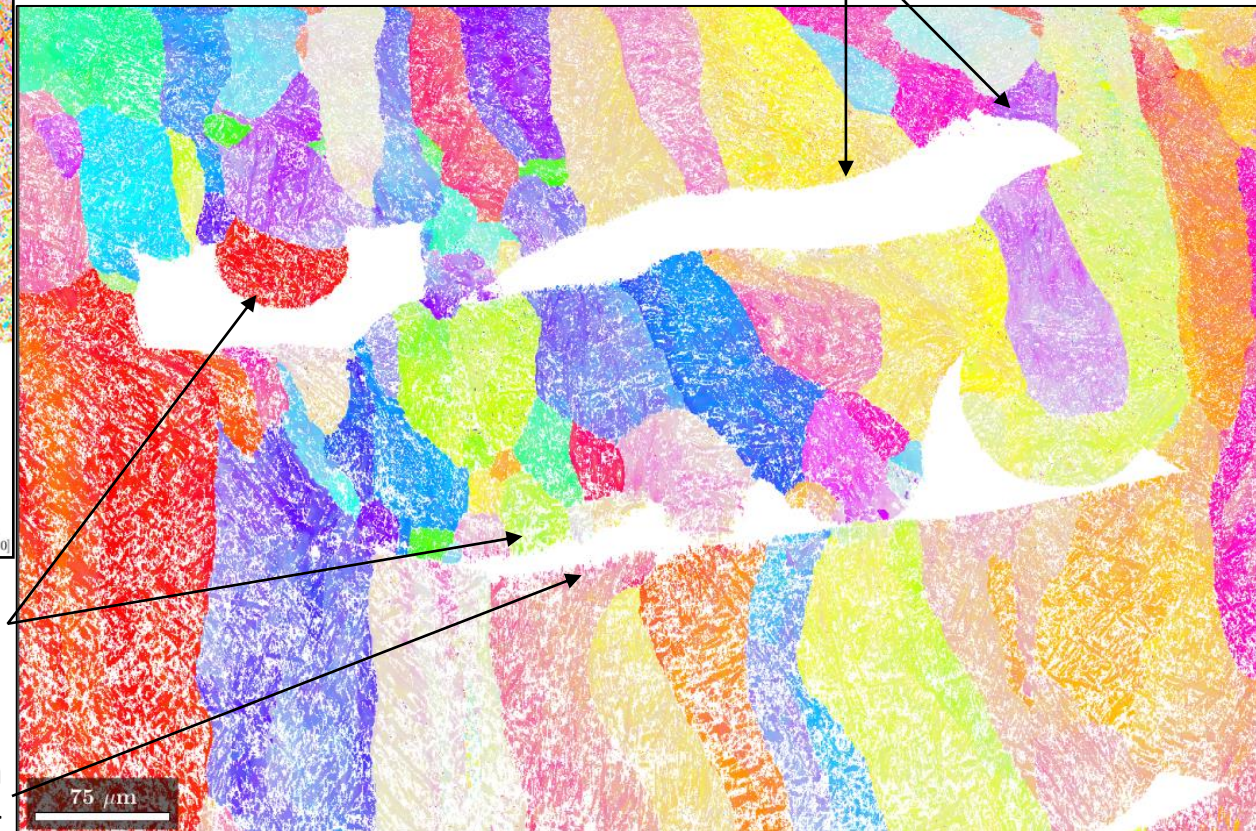
Parent Grain Reconstruction

MTEX/MATLAB and OIM Analysis (TSL)

Beta reconstruction helps us examine the effect of defects on grain growth and nucleation.



Continued growth past the defect

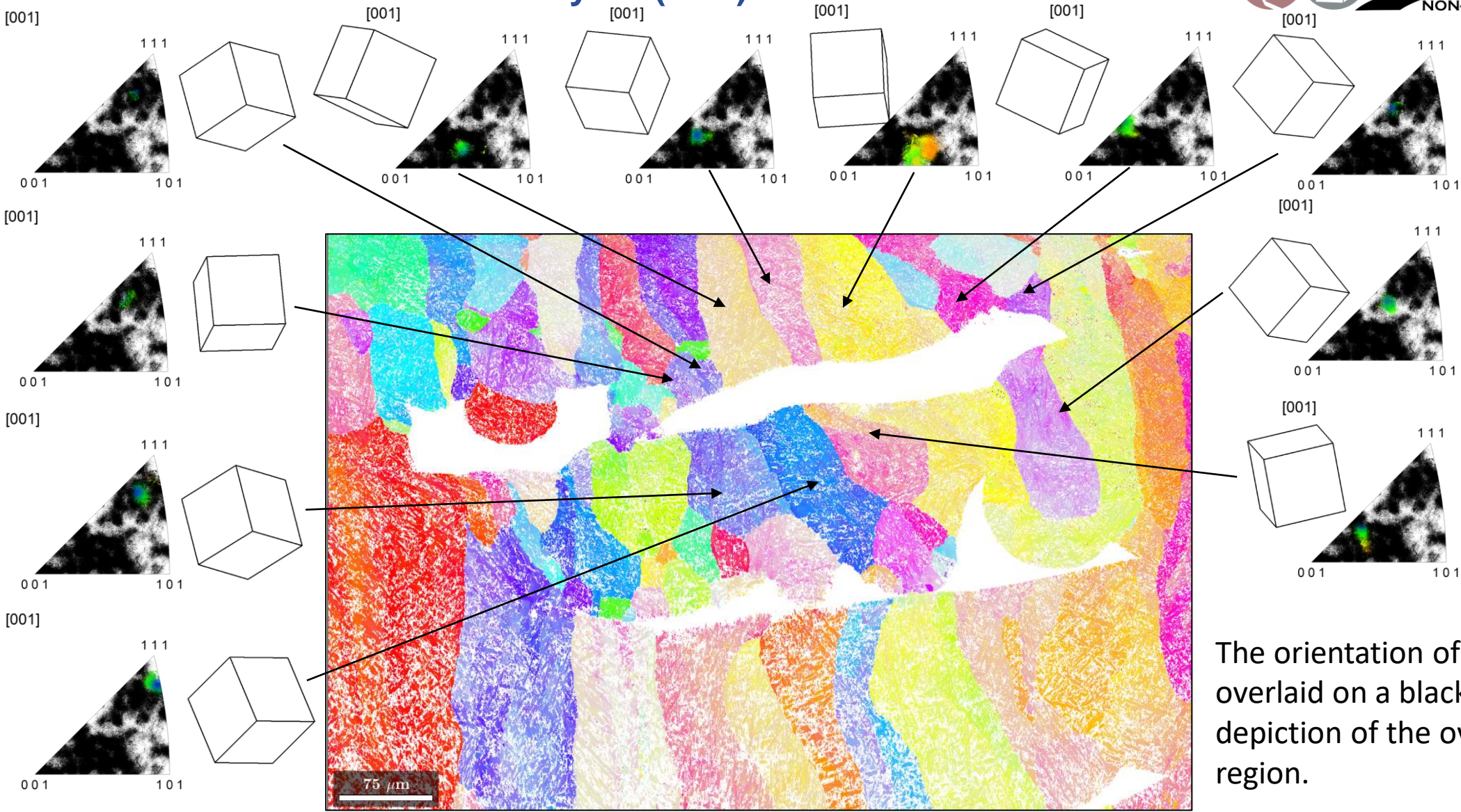


Lack-of-fusion defects can act as thermal barriers, preventing heat transfer. Overhanging powder particles create nucleation sites for new grains.

Nucleation of new grains
Columnar growth stopped by defect

Parent Grain Reconstruction

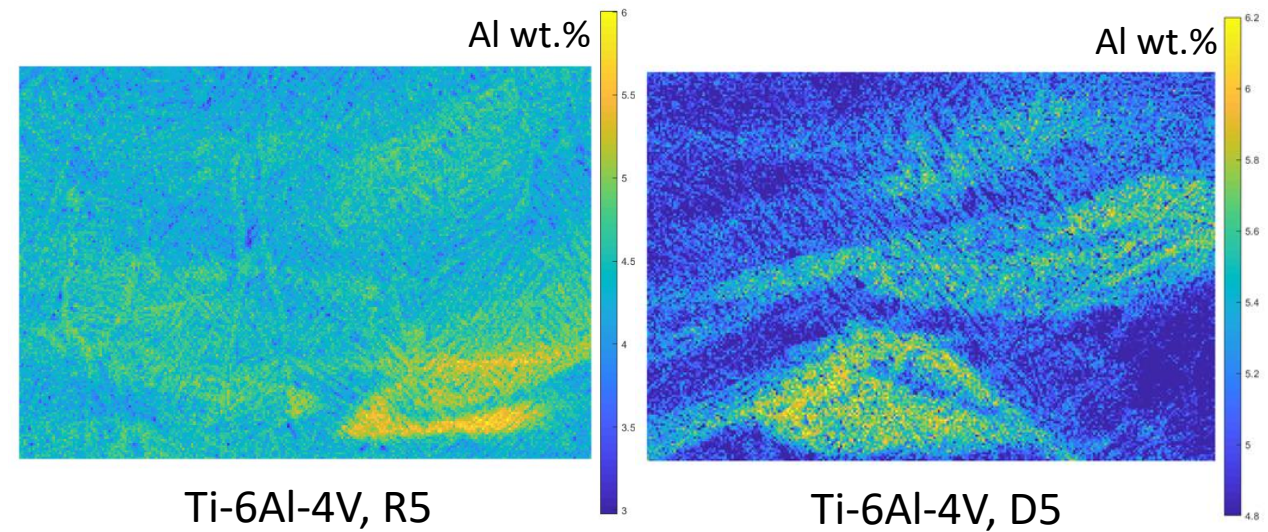
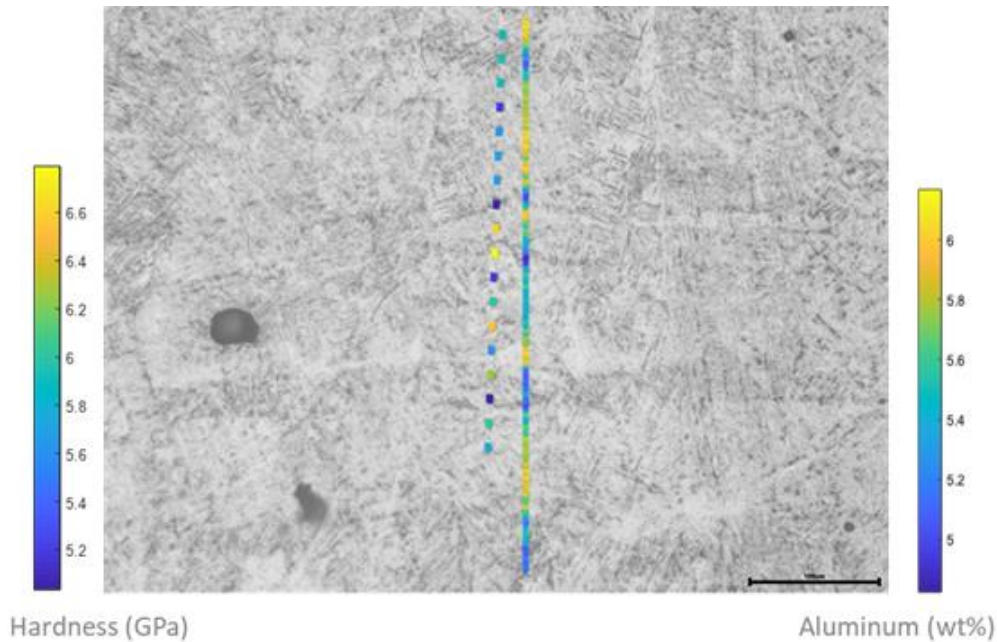
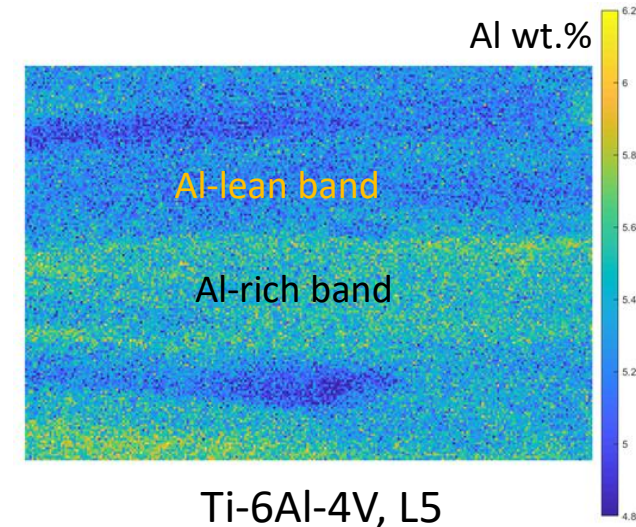
MTEX/MATLAB and OIM Analysis (TSL)



The orientation of each grain is overlaid on a black and white depiction of the overall texture of the region.

Local Property Variations

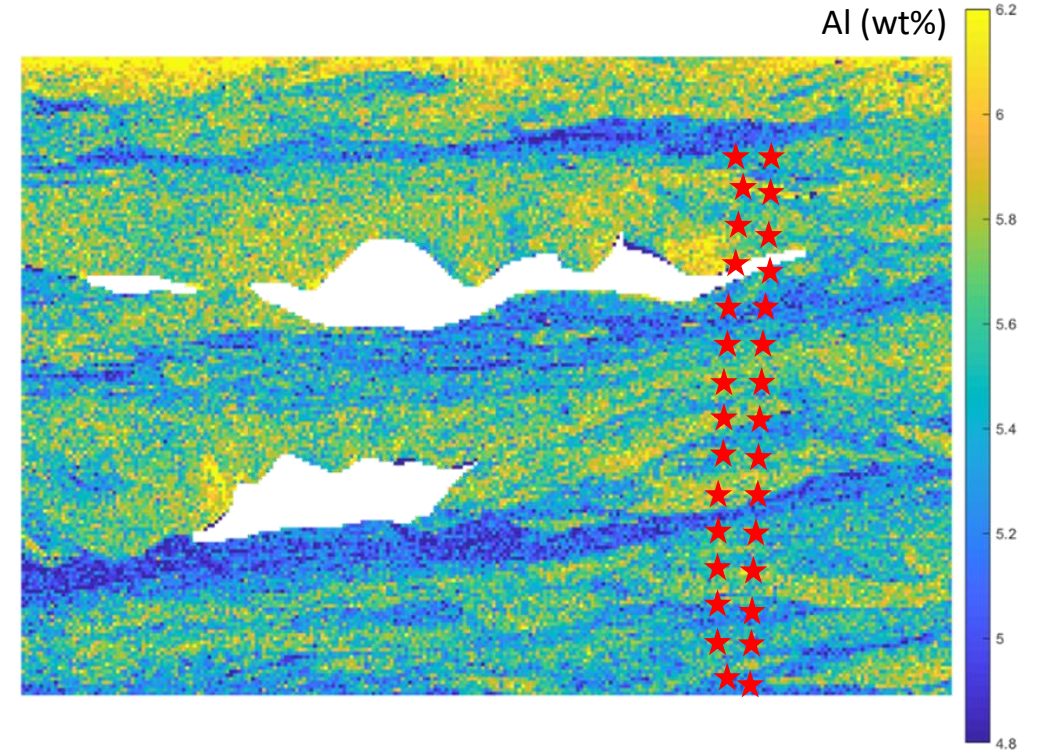
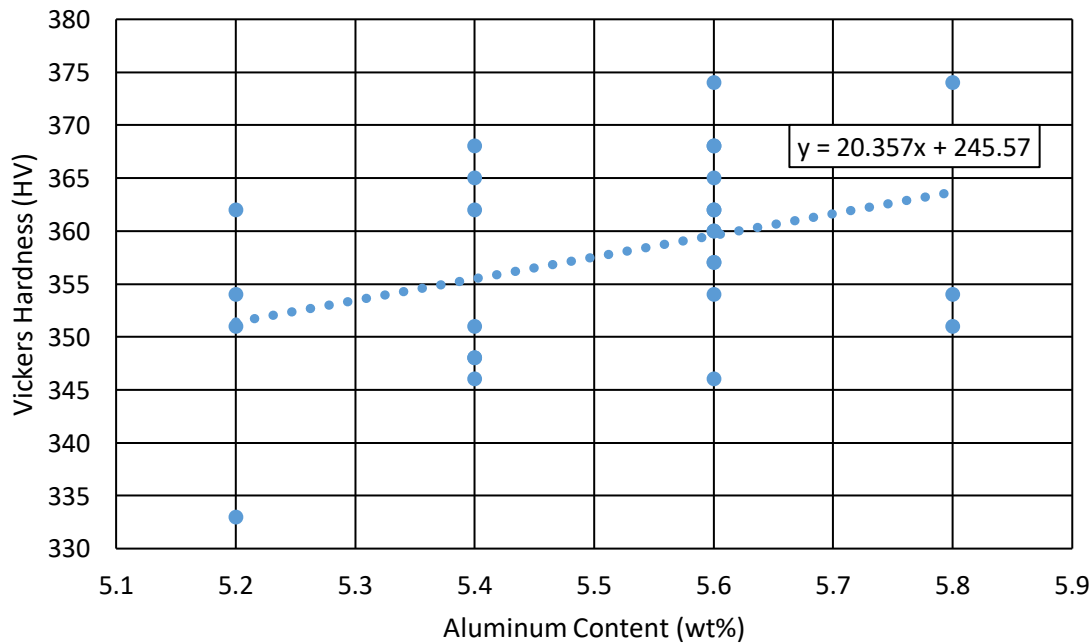
Previously, we have seen compositional variations across Ti64 scan strategies, in both EDS maps and through optical microscopy



Local Property Variations

MICROINDENTATION

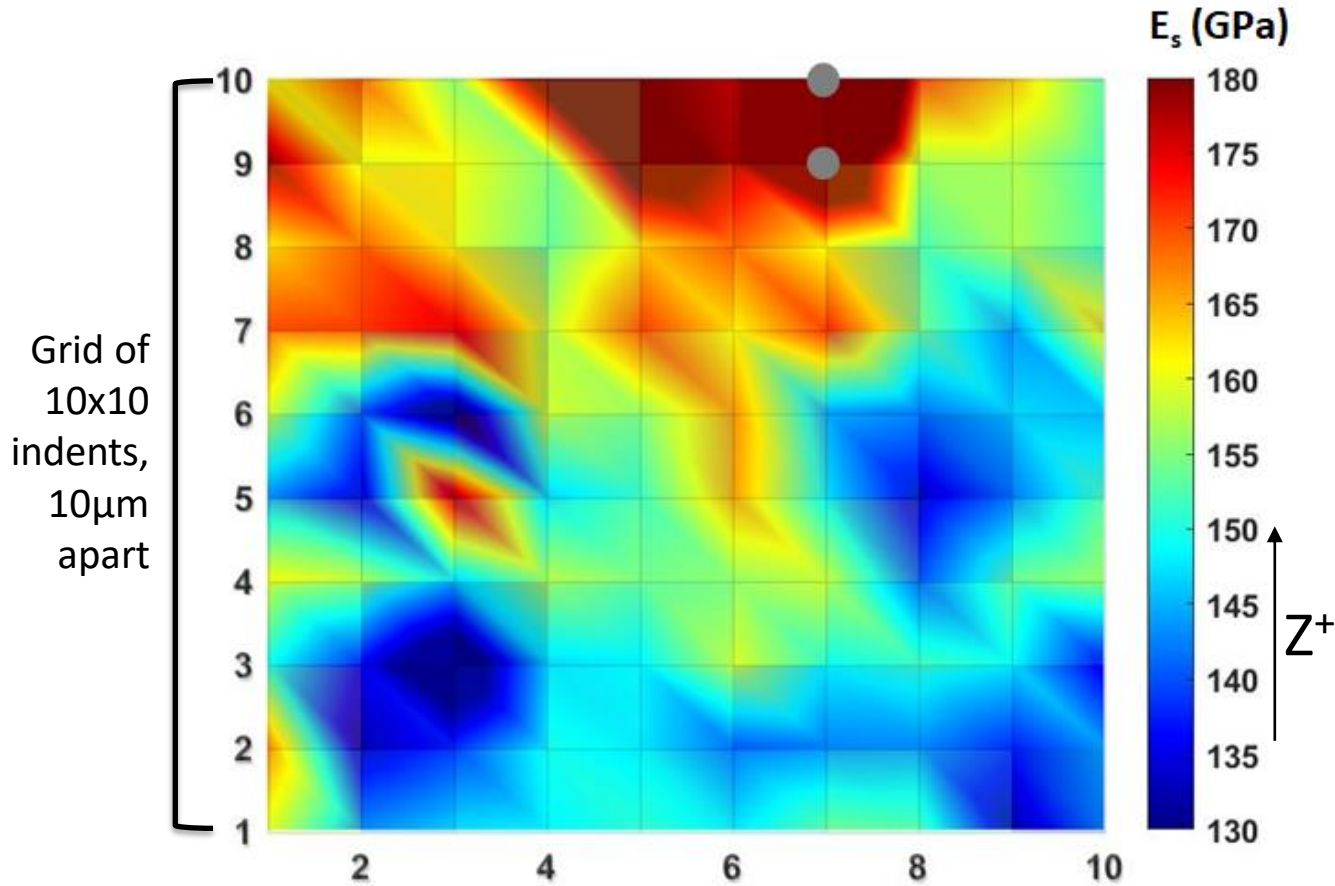
- Preliminary microhardness data suggests a correlation between local aluminum content and mechanical properties
- An estimate of aluminum content in the area of each indent suggests a general linear trend



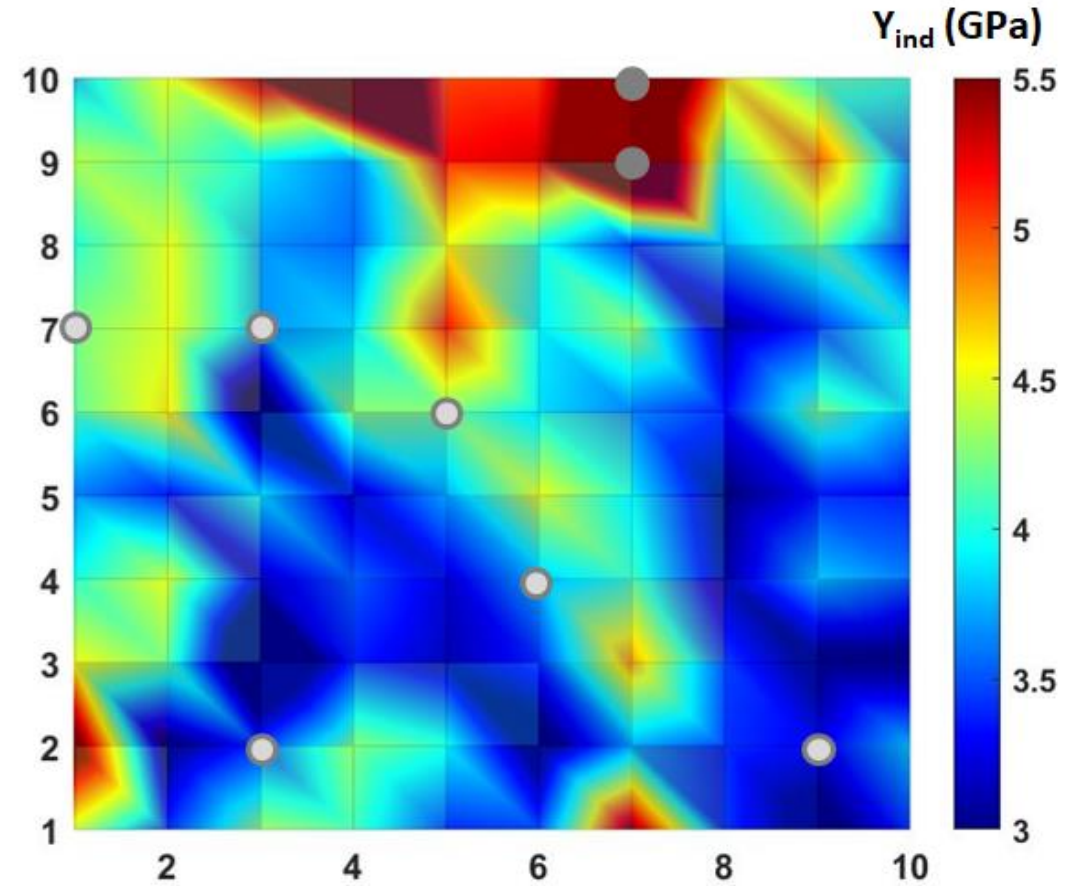
Local Property Variations

NANOINDENTATION

Effective Sample Modulus (GPa)



Indentation Yield Strength (GPa)

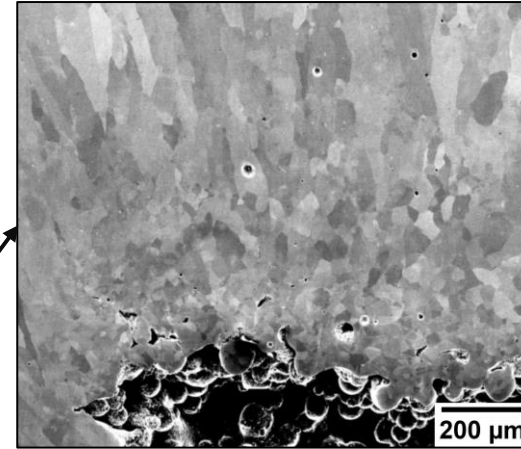
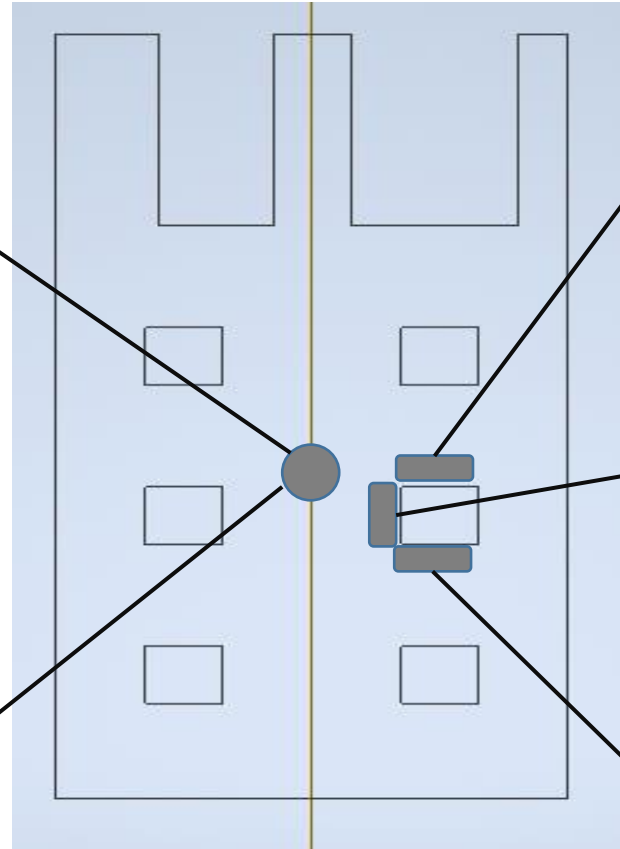
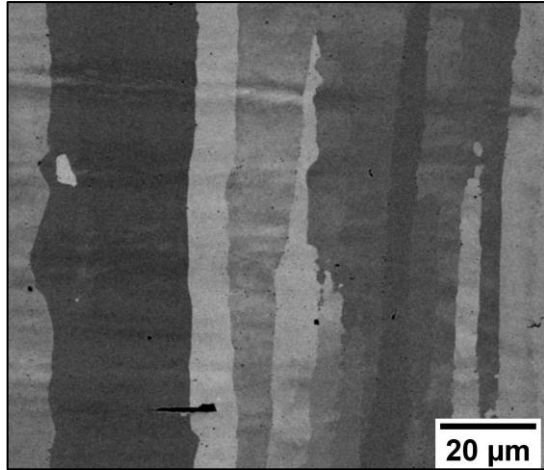


Banding observed in both OM and EDS maps is not clearly observed in these preliminary data.

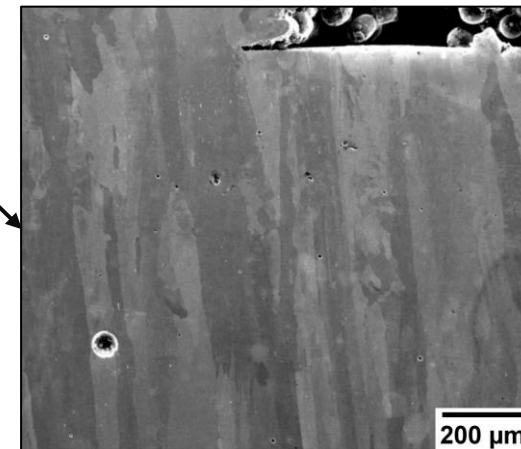
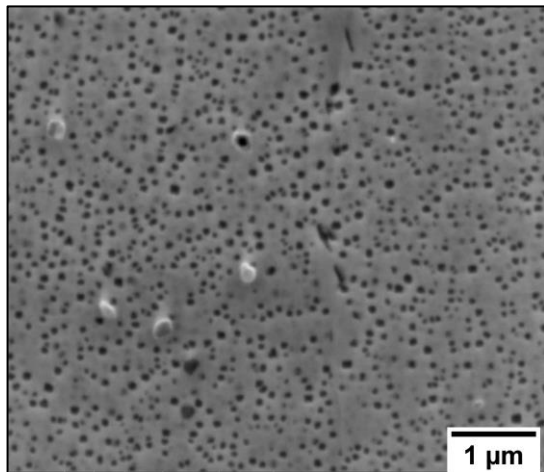
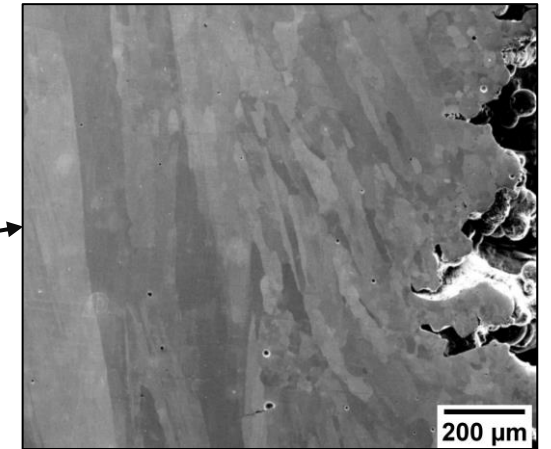
- Uncertain/bad modulus data
- Large pop-in (bad indentation yield data), interpolated values based on neighboring tests used for plot

Haynes Microstructure

Initial Observations



As expected, designed pores influence grain growth.



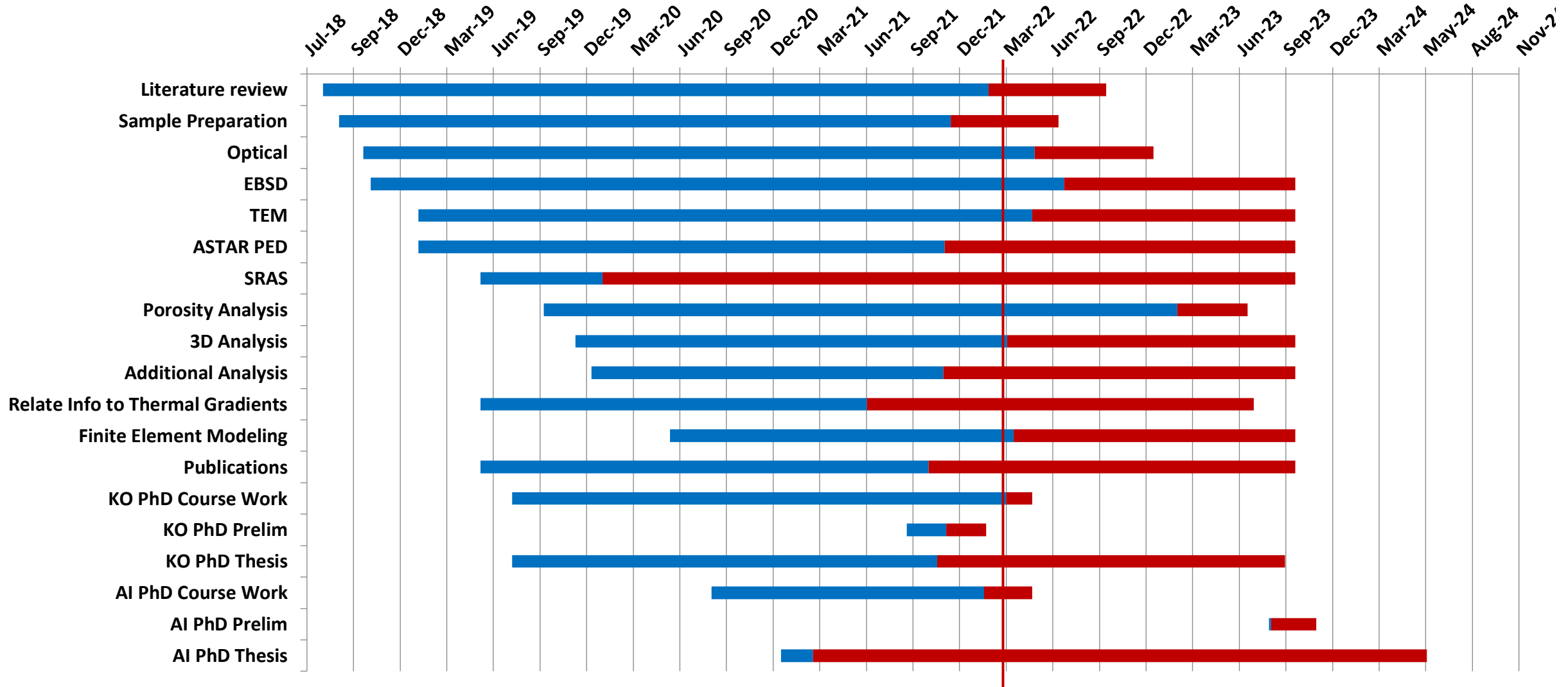
Future work:
Defect characterization, morphology, and distribution (pores, LOF, hot tearing)
Characterization of γ grain sizes and orientation/texture
Distribution and size of γ' throughout the build

Paper Updates



- *Differences in defect distribution across scan strategies in electron beam AM Ti-6Al-4V*
 - M.J. Quintana, K. O'Donnell, M. Kenney, P.C. Collins
 - Published; Advanced Materials & Processes, July-August issue
- *On the use of defects to understand physical processes in additive manufacturing*
 - K. O'Donnell, M.J. Quintana, M. Kenney, P.C. Collins
 - Submitted to Met. Trans. A.

Progress



Thank you!

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