



Sandia
National
Laboratories

SAND2020-5605C

Addressing Complexity in Additive Manufacturing

PRESENTED BY

Bradley Jared, Materials Engineering & Manufacturing S&T



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2 Complexity is Available

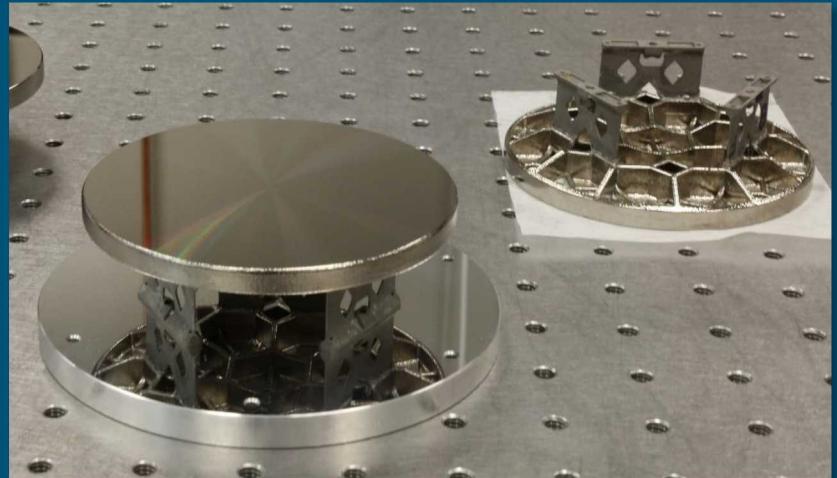
Advanced manufacturing provides ever expanding design freedom & value

- complex freeforms, internal structures, integration

Geometry

Engineered materials

- gradients & microstructure



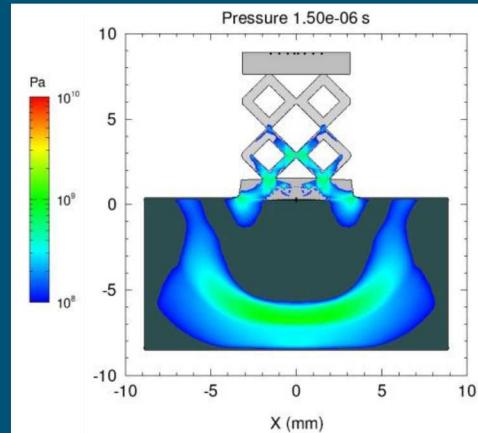
AM Ti6Al4V mirror w/diamond turned electroless Ni coating, Jared, ASPE, 2017



Photo by Mike Bejarano and Mark Olona, Sandia Lab News



316L SS HOT SHOT top cap on plate



CTH simulation of 250m/s lattice impact



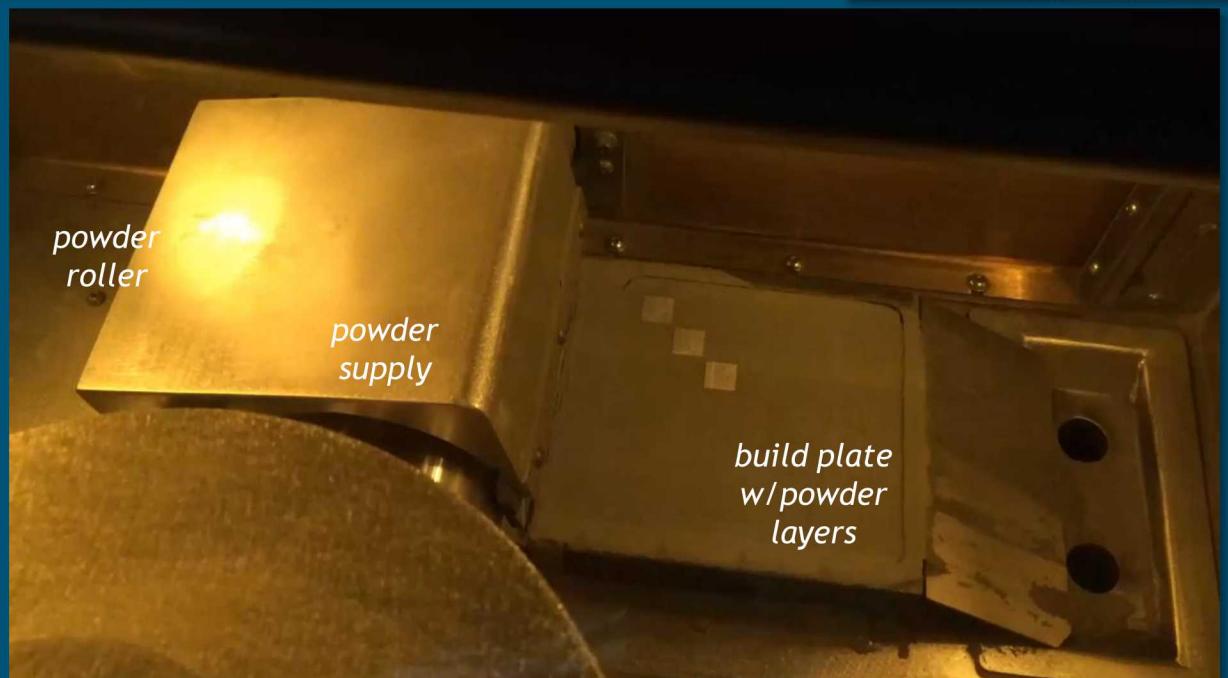
3 Laser Powder Bed Fusion

3D Systems ProX 200

- CW ytterbium, single mode fiber laser
- 1070nm, 300W max power, 100 μ m dia.
- scan speed = 1.4m/sec
- rotating powder roller
- Ar backfill & cover gas flow

Part capabilities

- 140x140x100mm build volume
- 316L stainless steel
 - 20-30 μ m diameter powder
- deposition rate \sim 100mm³/min
- 10⁵-10⁶°C/s heating & cooling rates



But, Complexity Isn't Free

Features tied to requirements incur costs

AM material formation concurrent w/geometry

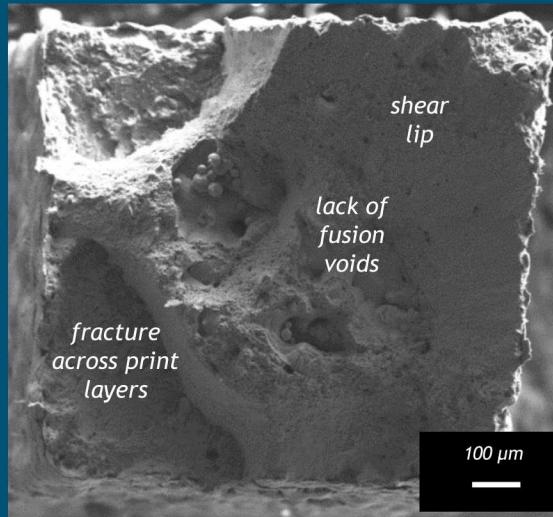
- want to predict part/material performance
- **how to ID a bad part?**
- significant design margins and/or rigorous post-process inspection / validation

Understand mechanistic impacts on properties

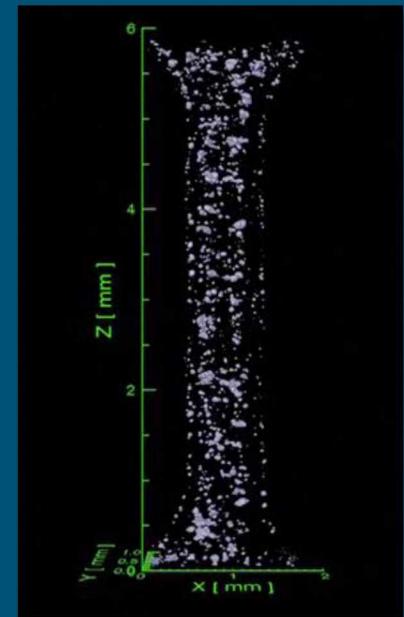
- build process-structure-property relationships to predict **margins & reliability**
- characterize stochastic response to design for **uncertainties**
- provide scientific basis for qualification of AM metals for high consequence applications

Quantify **critical** material defects & **useful** signatures

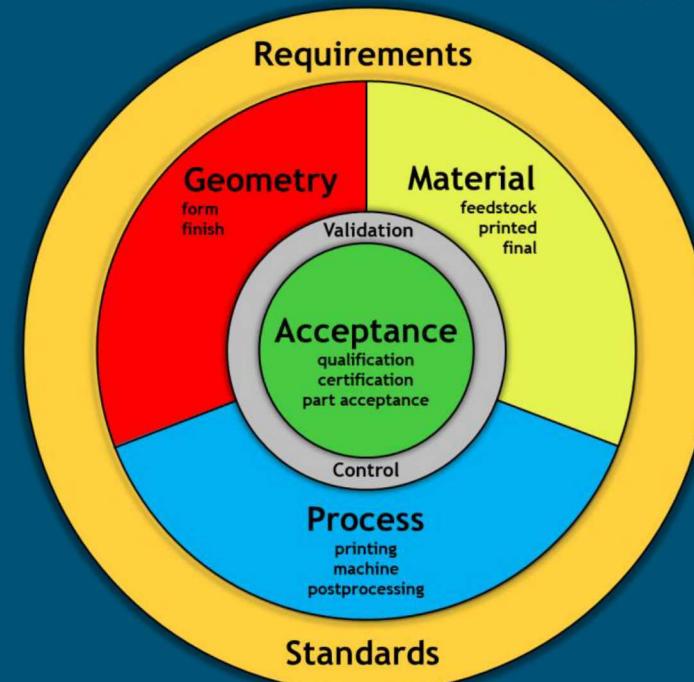
- D-tests, NDE, process monitoring, mod-sim



17-4PH dogbone fracture surface



17-4PH dogbone porosity



*elements of qualification
Jared, Sci Tech and Appl of Metals in AM, 2019*



Lack of Process Control Produced Material Uncertainty in Early AM Metals



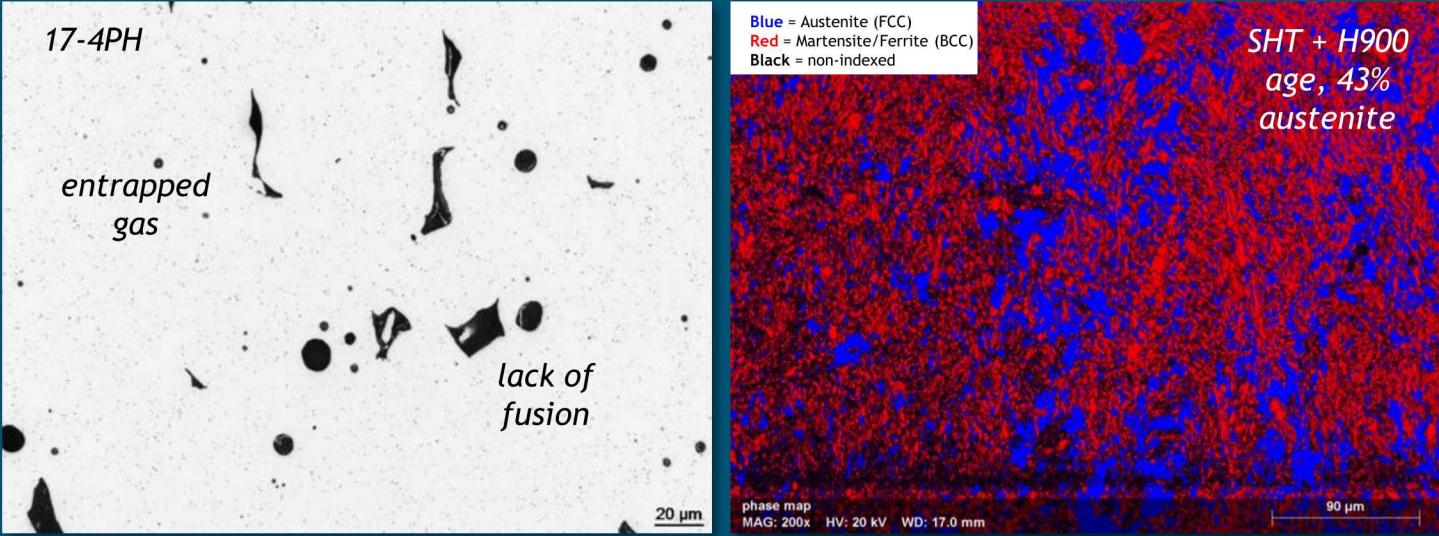
17-4PH parts from external vendor

- analysis confirmed 17-4PH composition, but unexpected microstructure

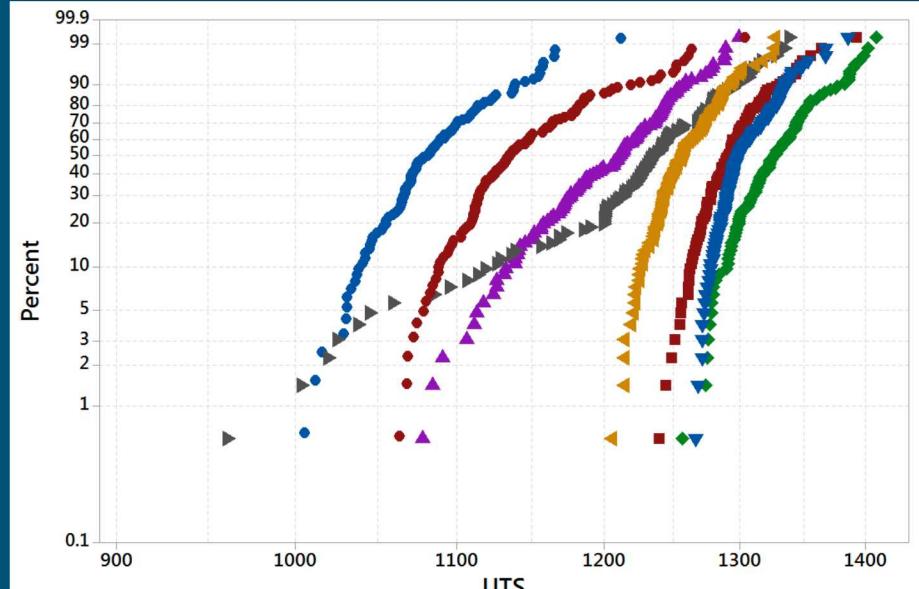
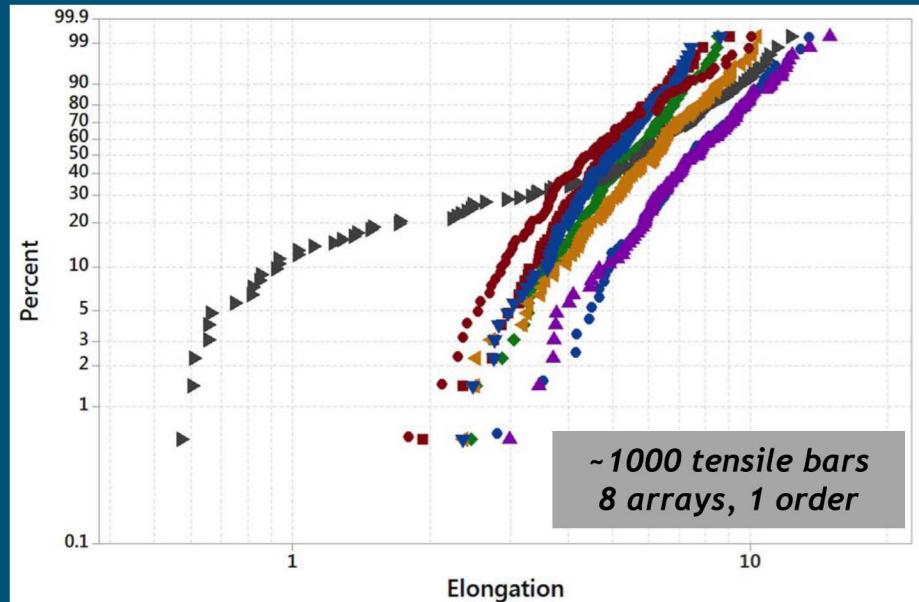
Sandia did not know

- feedstock pedigree, machine, build environment, process inputs or post-processing steps

Not-surprisingly, extensive material variability observed



Jared, SAND2017-13190



High Throughput Tensile Testing (Gen I)



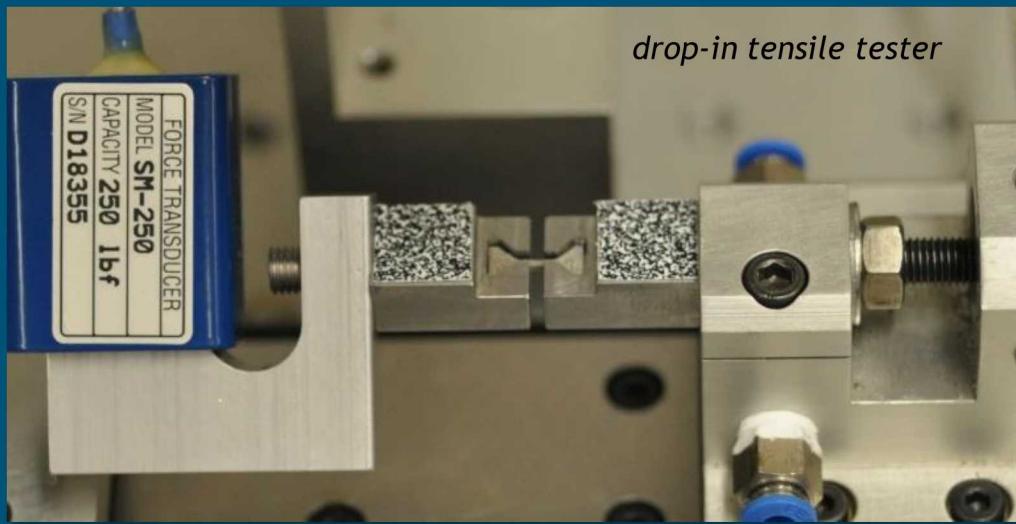
Monolithic build w/110 dogbones

- custom design per ASTM

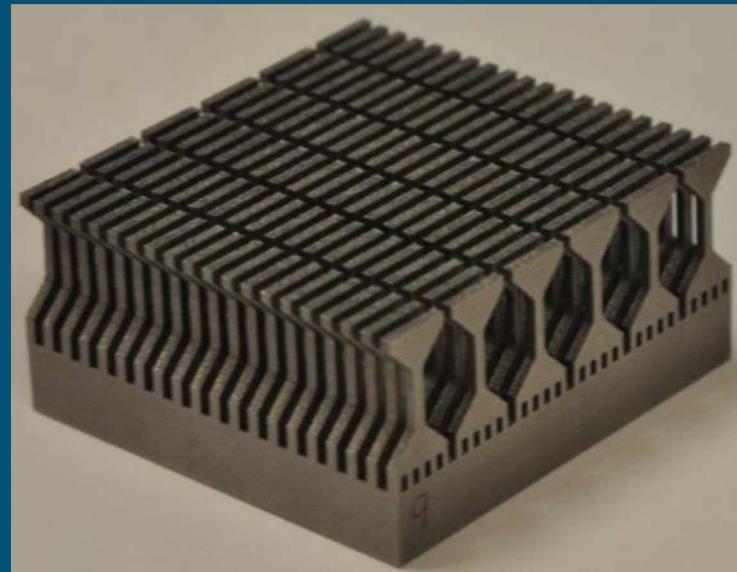
Digital image correlation (DIC)

Necessary to rapidly capture material distributions

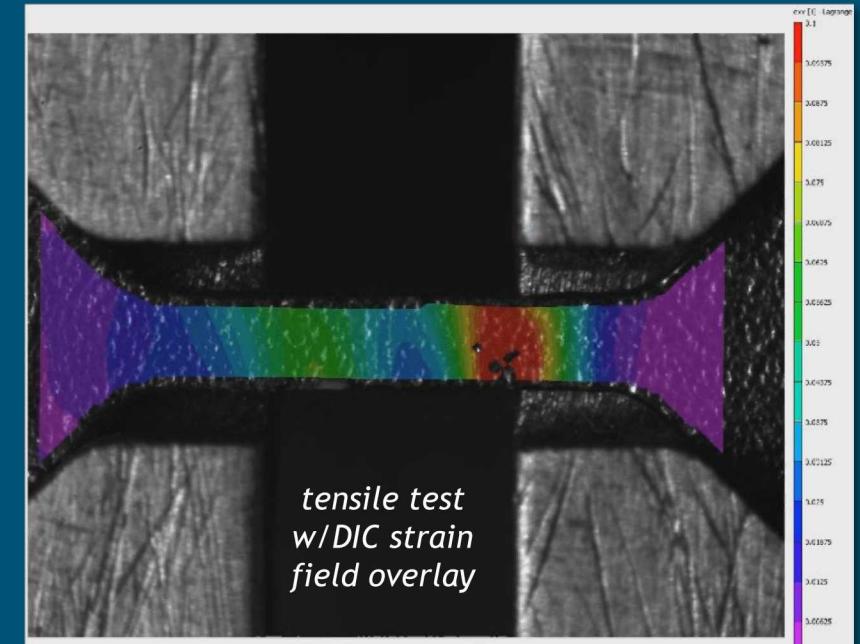
- applicable for the lab & production



drop-in tensile tester



high throughput test sample w/ 120 dogbones,
1x1mm gage x-section



tensile test
w/DIC strain
field overlay

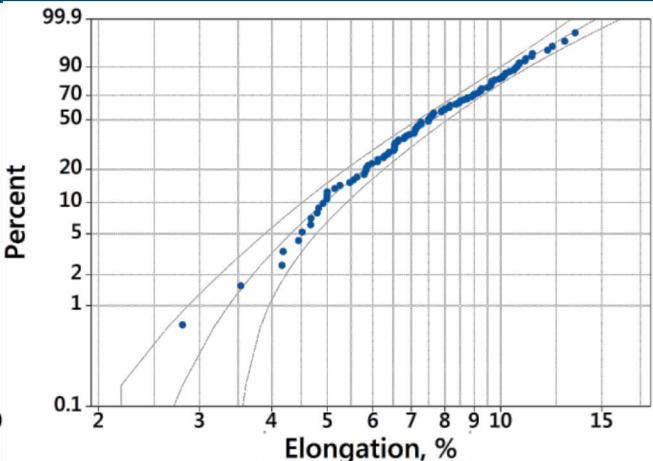
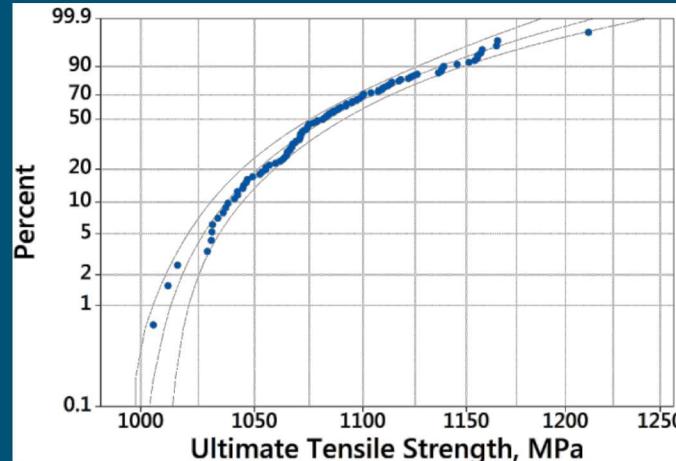
Salzbrenner, *Journal of Materials Processing Technology*, 2017

Stochastic Response of AM Metals

Defect dominated failure

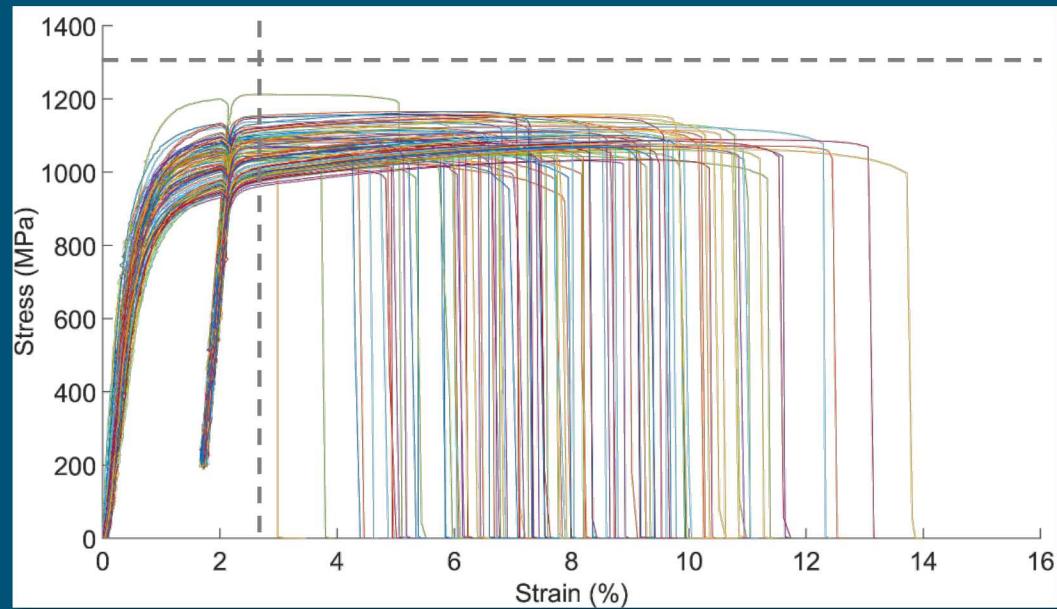
- 3-parameter Weibull informs design threshold
- ductile dimples & shear rupture planes
- voids & lack-of-fusion boundaries are likely crack nucleation sites

How to capture efficiently & accurately?

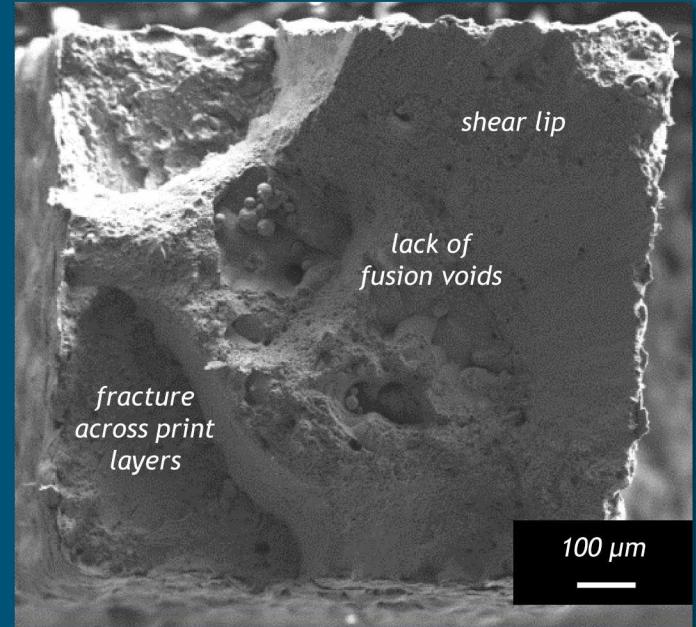


material performance fit to 3-parameter Weibull distributions

AMS spec for H900: modulus = 197 MPa, yield = 1172 MPa, UTS = 1310 MPa, strain at failure = 5%

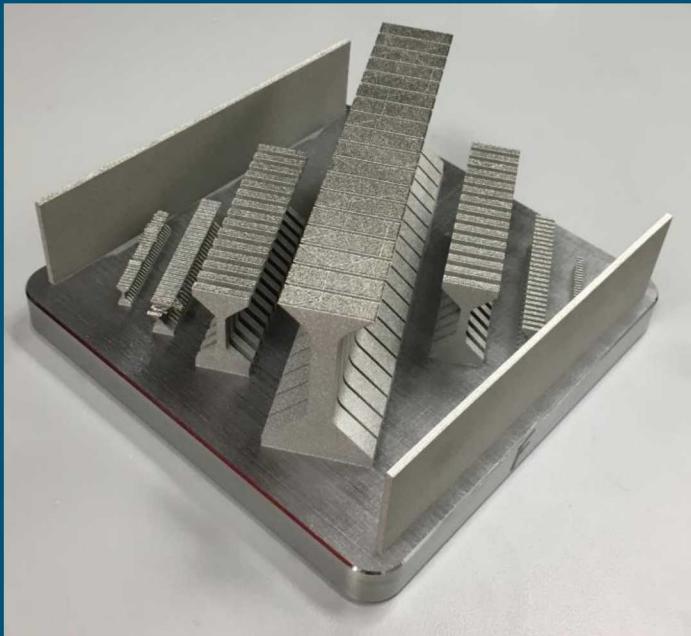


110 stress-strain curves for 17-4 PH after SHT+H900

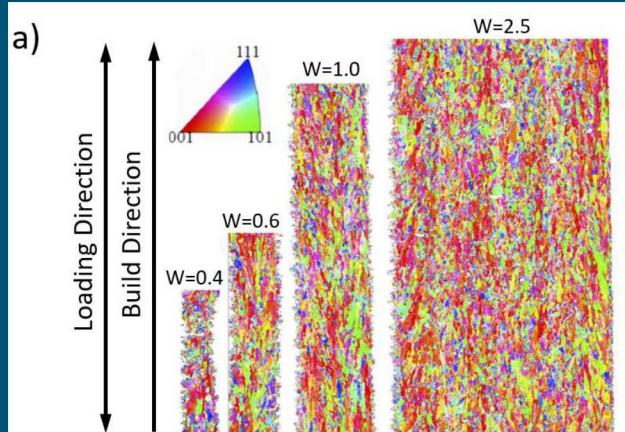


failure at 2% elongation, SHT+H900

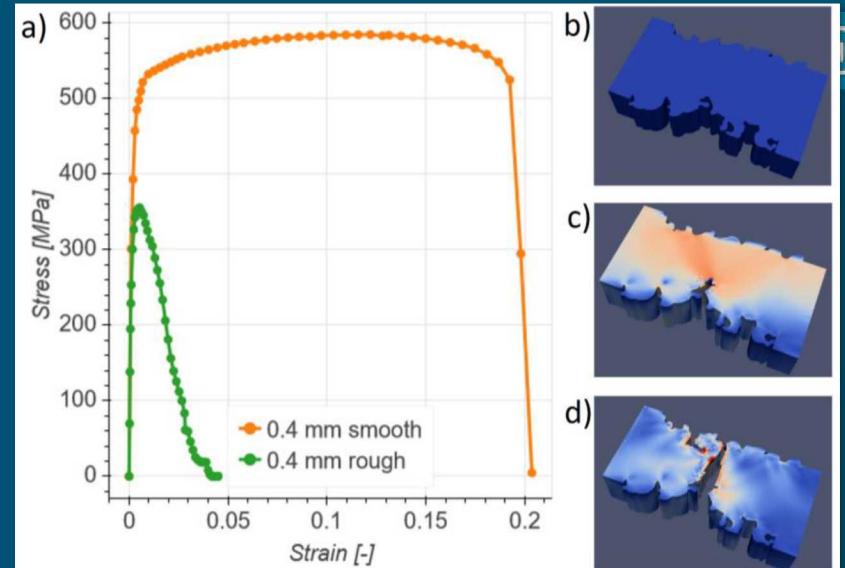
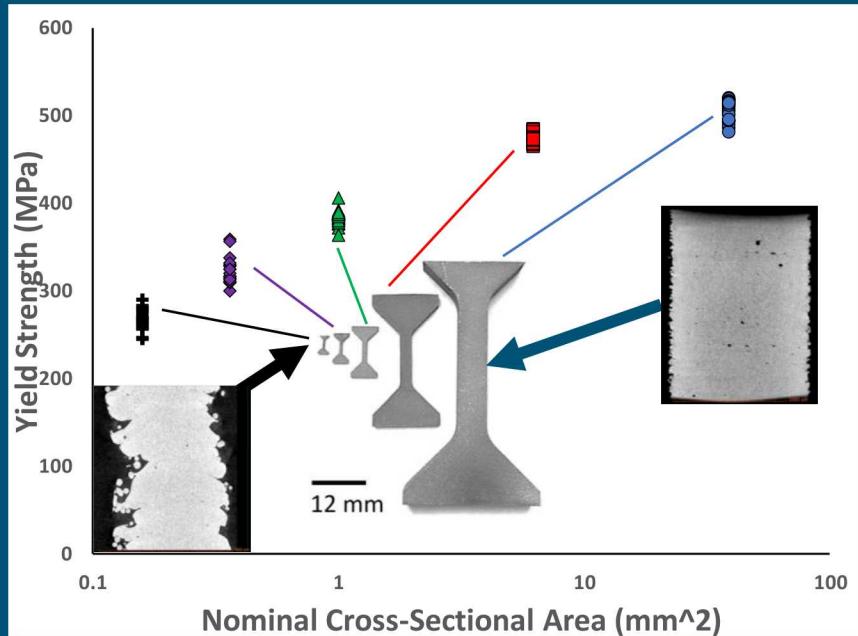
Size Effects



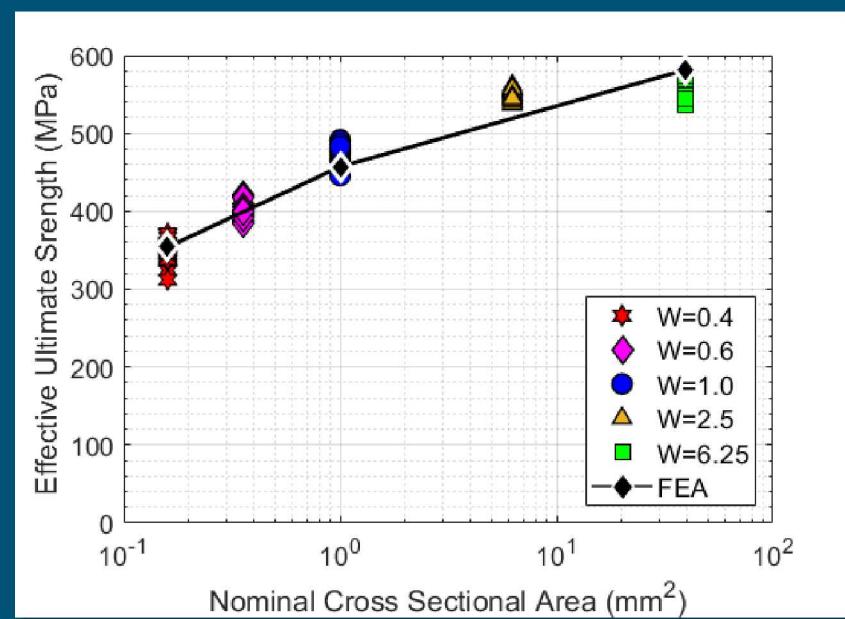
Prox 200 build plate w/tensile samples



microstructure is invariant w/sample size



surface roughness dramatically reduces strength of features with similar size scale



FEA model incorporating surface finish correlates well to experiments

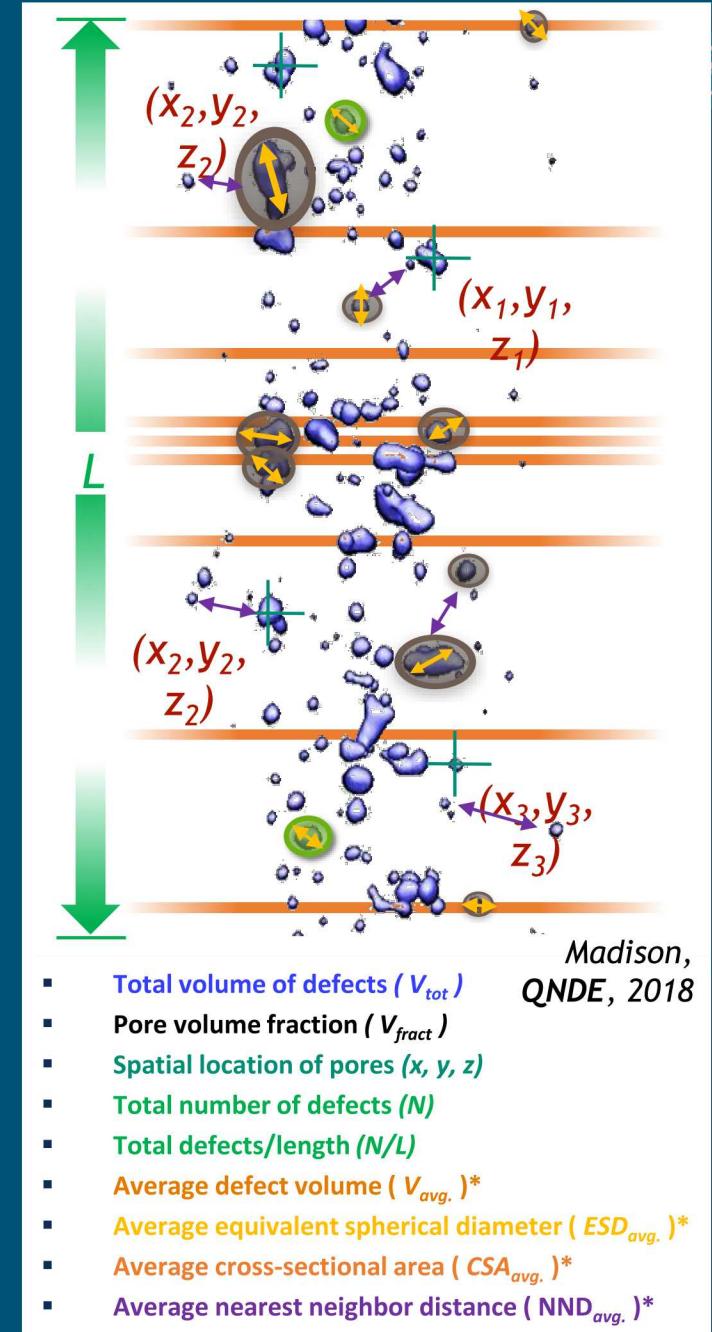
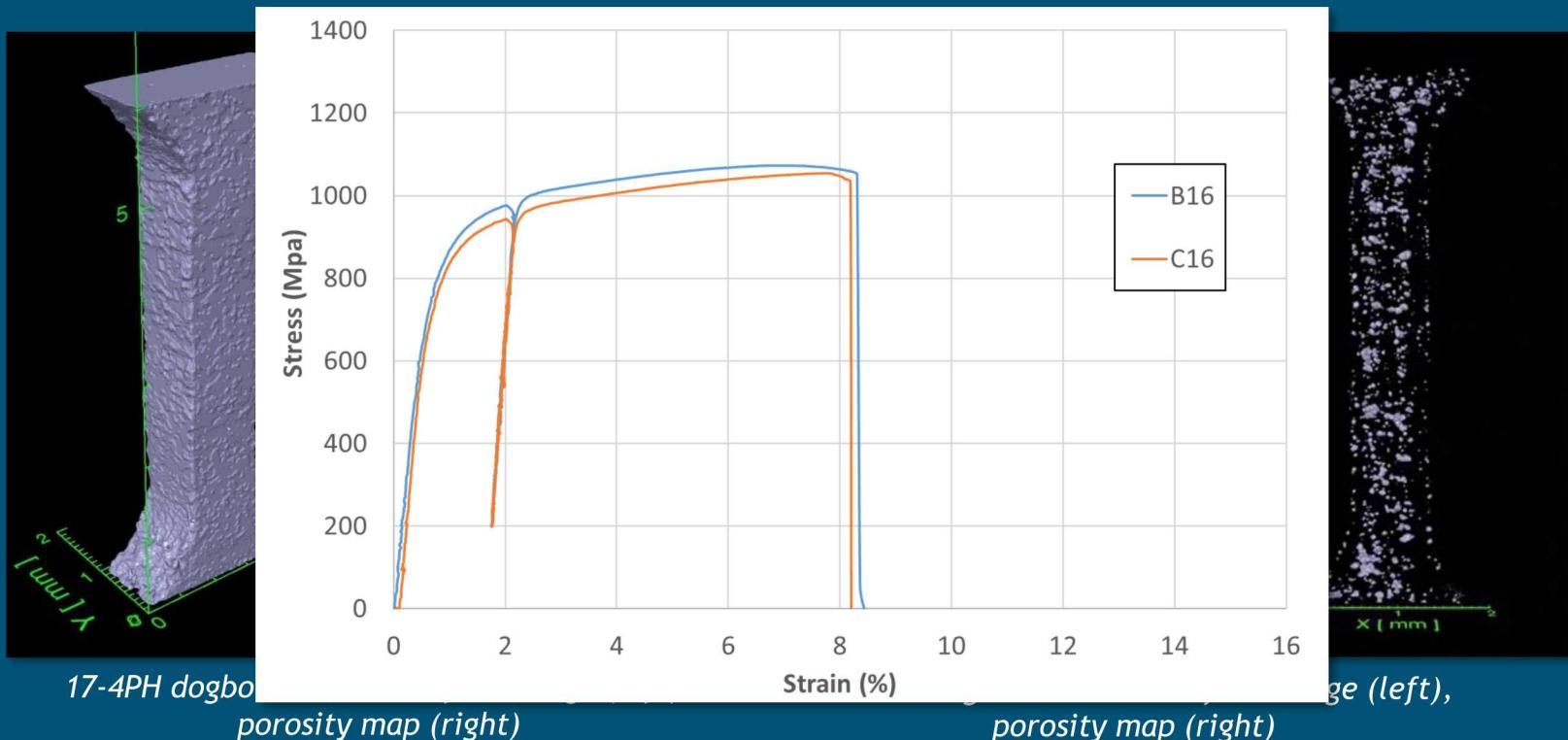
Tying Structure (Porosity) to Properties

Extensive work using computed tomography (CT)

- multiple potential metrics exist
- correlations are immature

Prediction of uniformly dispersed porosity fields is difficult

- behavior falls within an expected performance distribution



Gross Defects Drive Performance Outliers

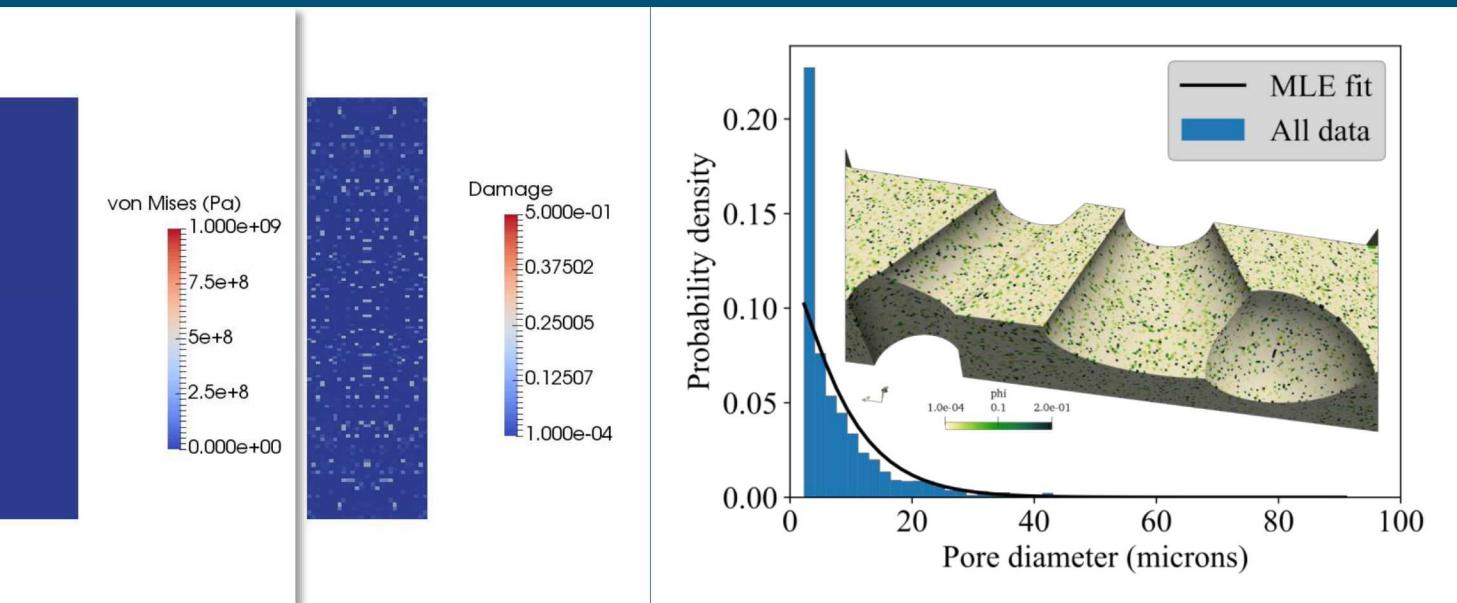
Failure initiates near large pores

Potential tolerance bound

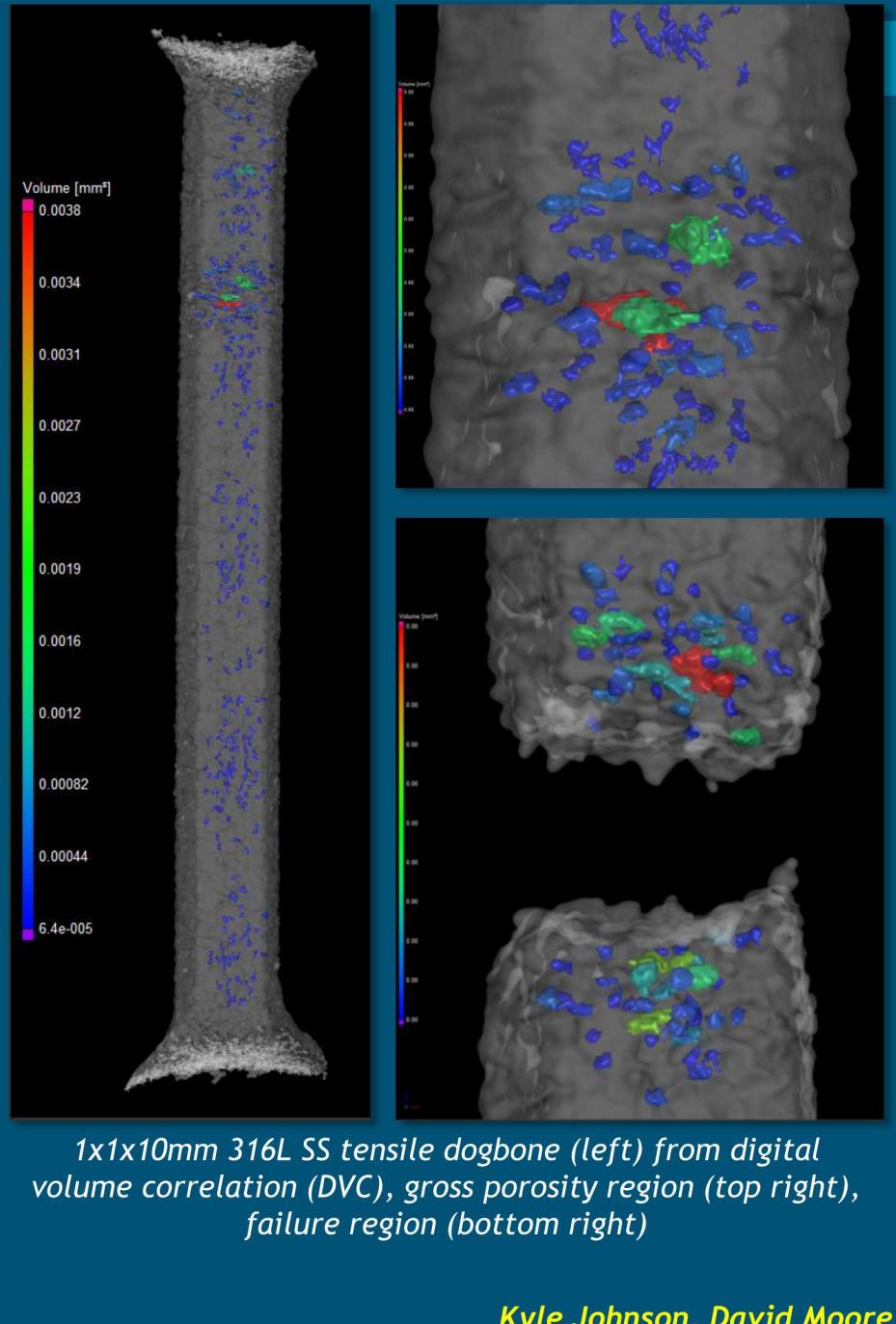
- design requirements
- CT inspection
- process monitoring

Performance simulation represent power analysis & diagnostics tools

representing porosity as initial damage



Johnson, Int J Fract, 2019



Monitoring Powder Reuse

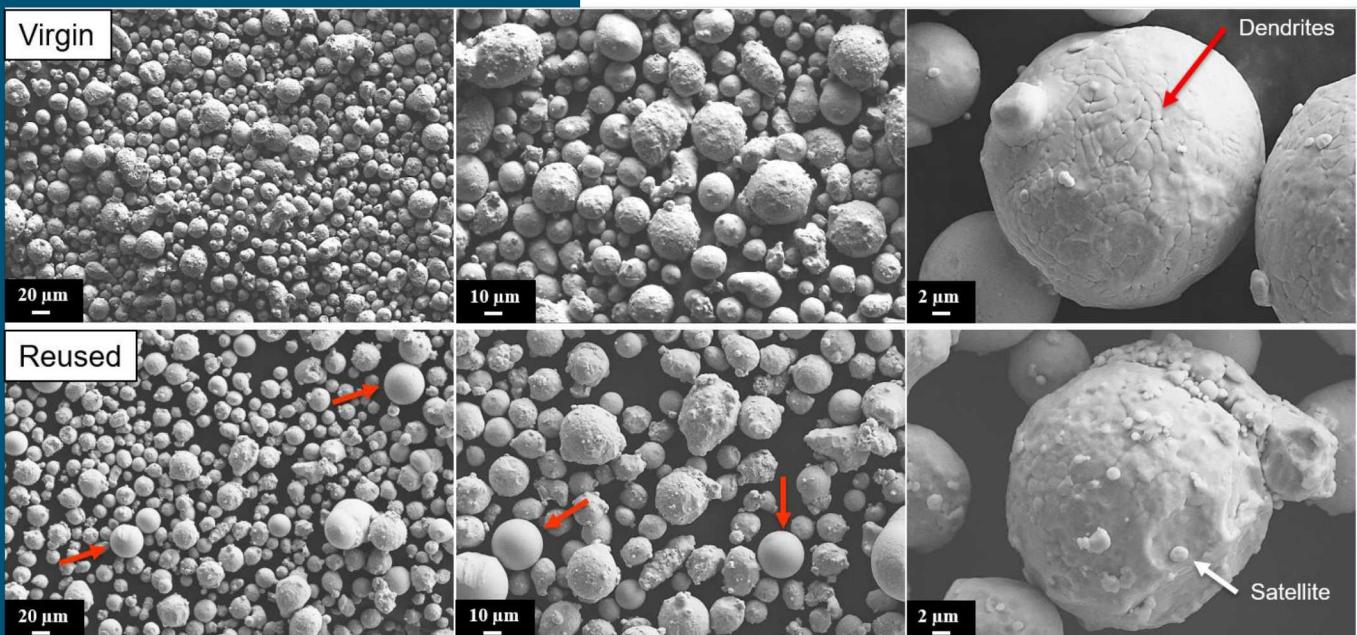
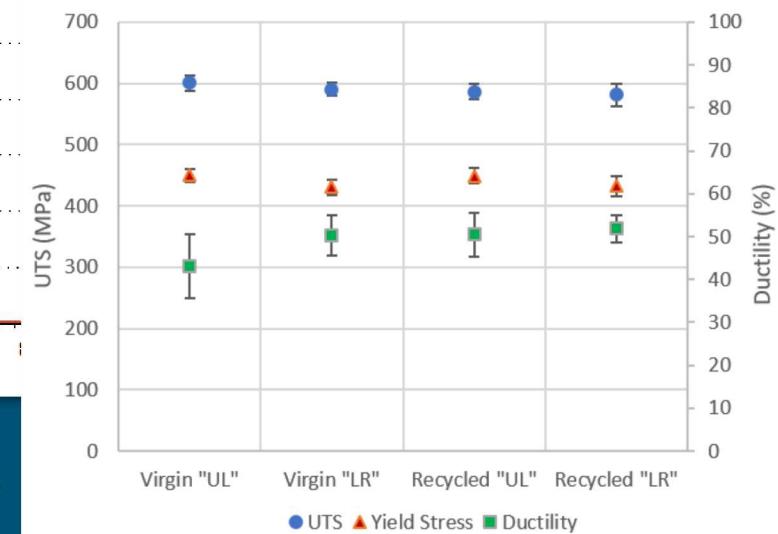
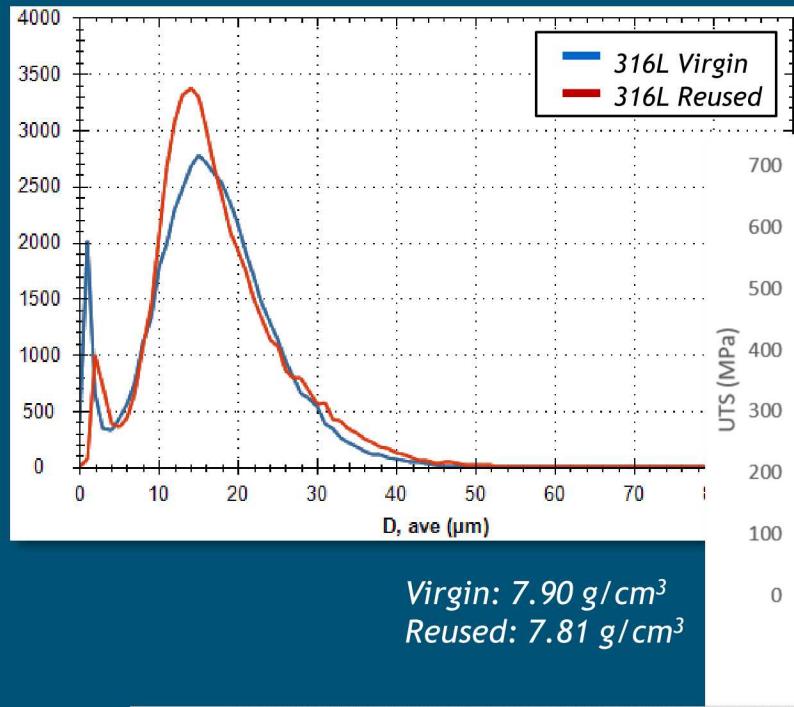
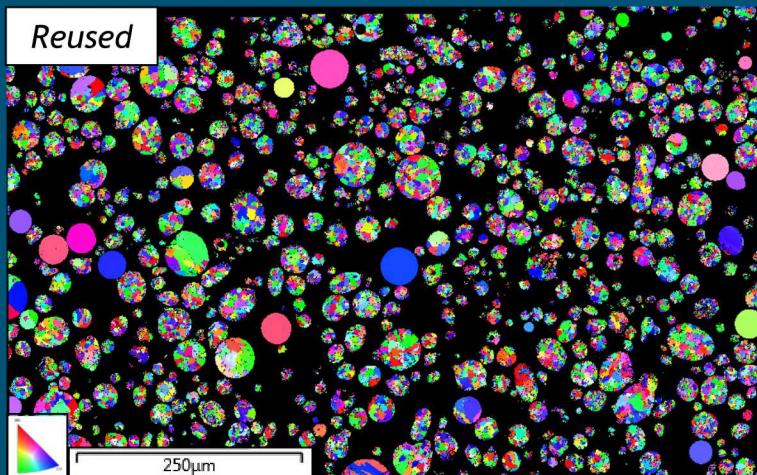


Tracking powder size, morphology & EDS composition w/reuse

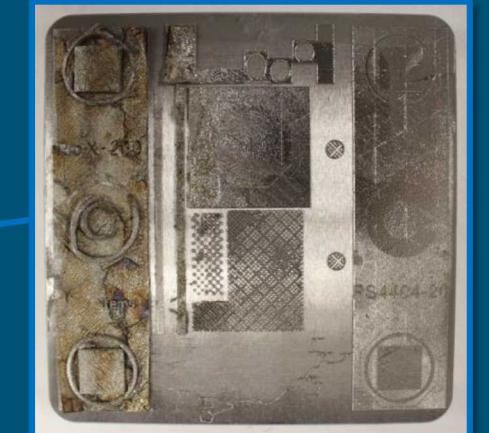
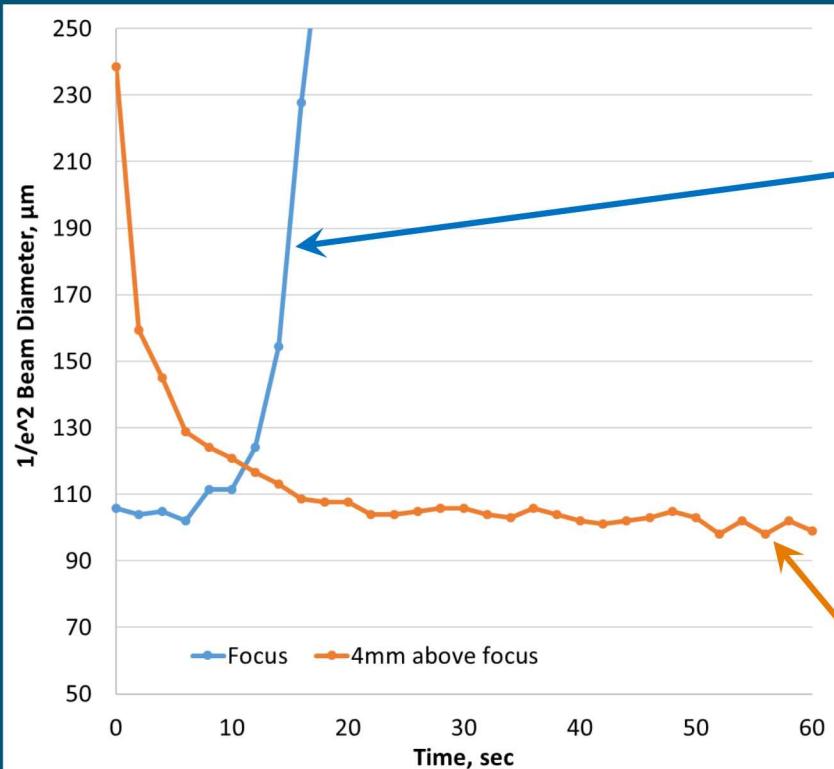
- satellites & agglomerates increase
- observe highly spherical, ferrite particles
- increase in fines & reduction in larger particles
- collected over 30 reuses w/powder under Ar

Material properties remain stable

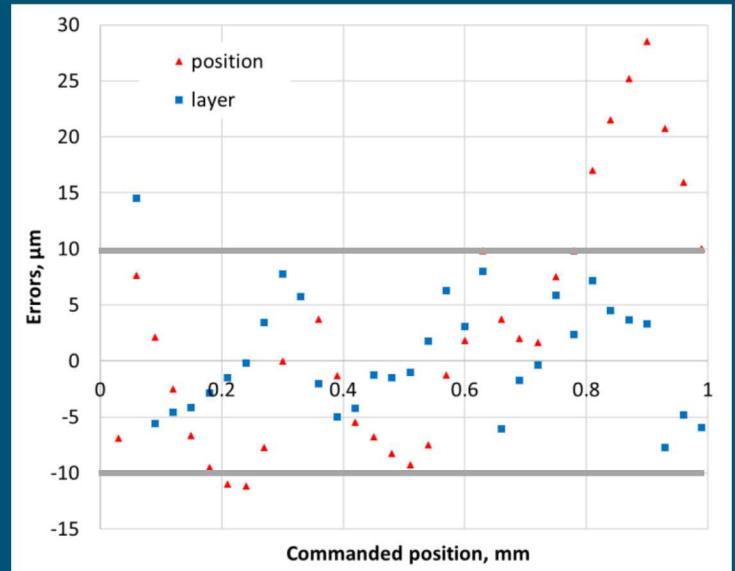
316L SS is a robust material for processing & properties



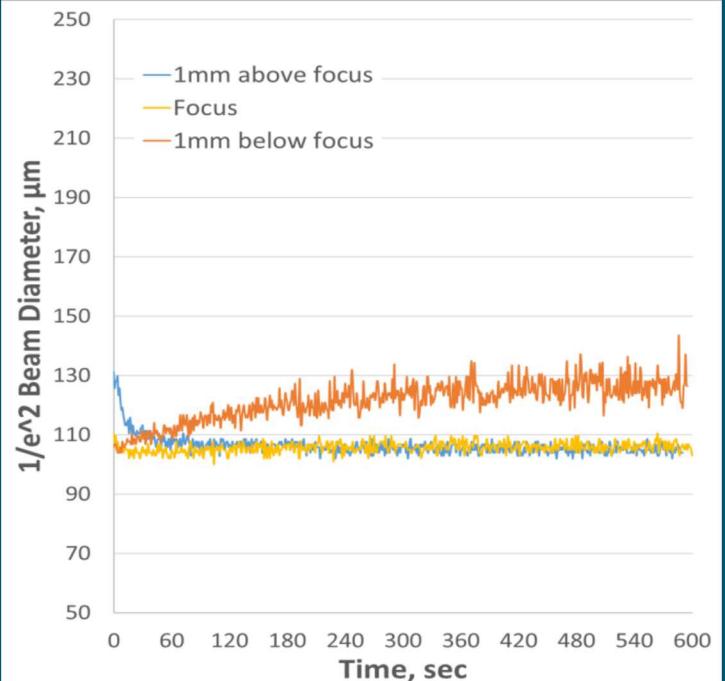
Machine Metrology is Critical to Assure Part Quality



layer thickness errors



*upgraded *f-theta* lens response*



Process-Structure-Property Process Maps



Parameters

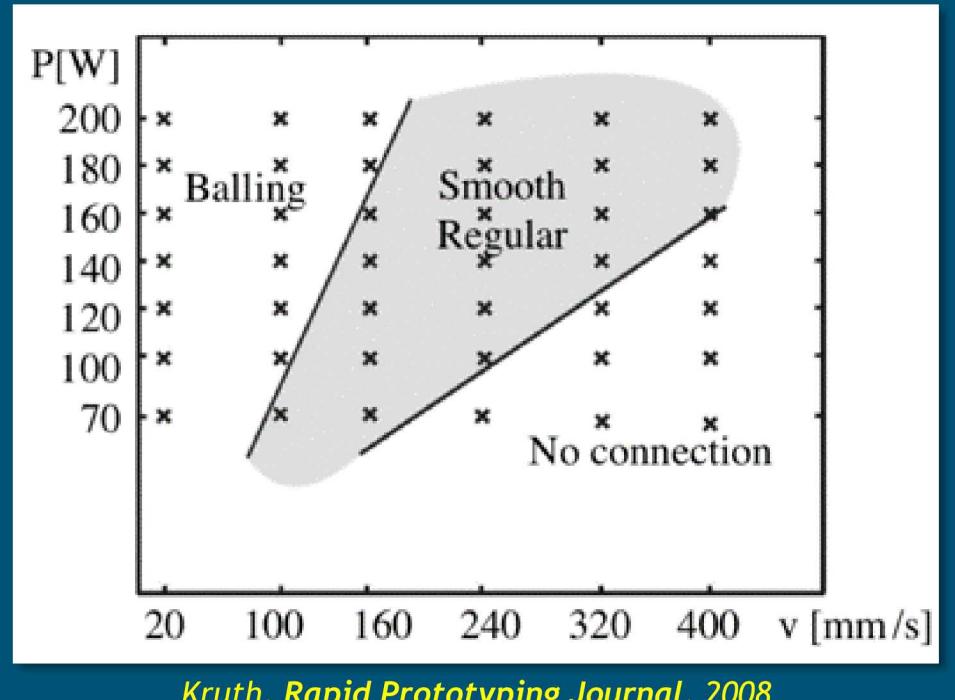
- 316L stainless steel
- laser power: 10-240W
- velocity: 50-2800mm/sec
- layer thickness: 30, 40 μ m
- average powder diameter: 15, 25 μ m
- laser focus offset: -1.5mm below focus to +3.5mm above

Experiment forms

- line scans, area pads, density cubes, HTT tensile, Charpy

Performance metrics

- surface finish, form error, density, tensile properties, Charpy toughness, microstructure
- where are optimal process settings?



Kruth, *Rapid Prototyping Journal*, 2008

The Influence of Process Variables on Physical and Mechanical Properties in Laser Powder Bed Manufacturing

- Josh Koepke, MS Thesis, UNM Dept of ME, 2019

Line Scans

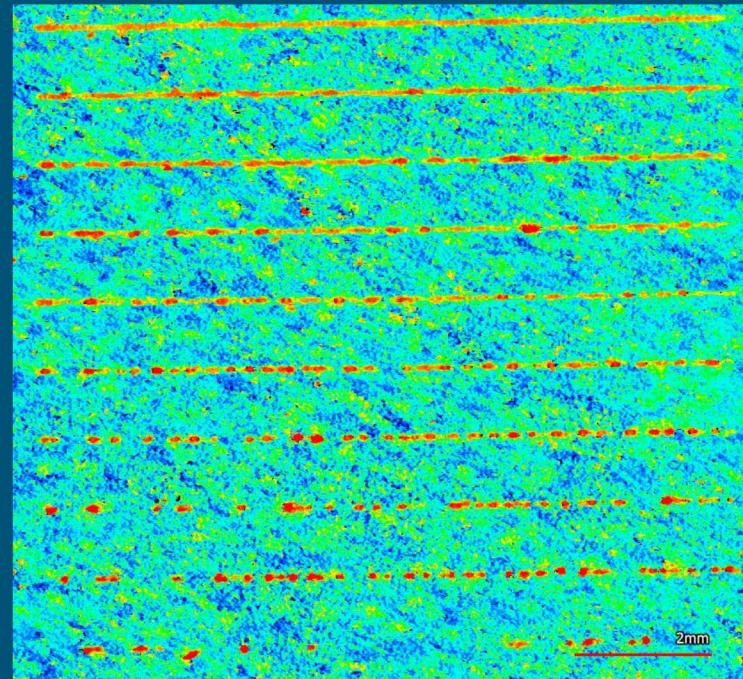


Substrates: bare plate, powder layer on 20 layer AM pad

- 60 lines on each substrate, 1.0 cm long
- power: 25-175W
- velocity: 250-2500mm/sec

Simplistic first step, but quick & informative

- capturing melt pool geometry via metallography
- useful to define nominal process boundaries for any material
- used to establish relevance of simple Rosenthal model

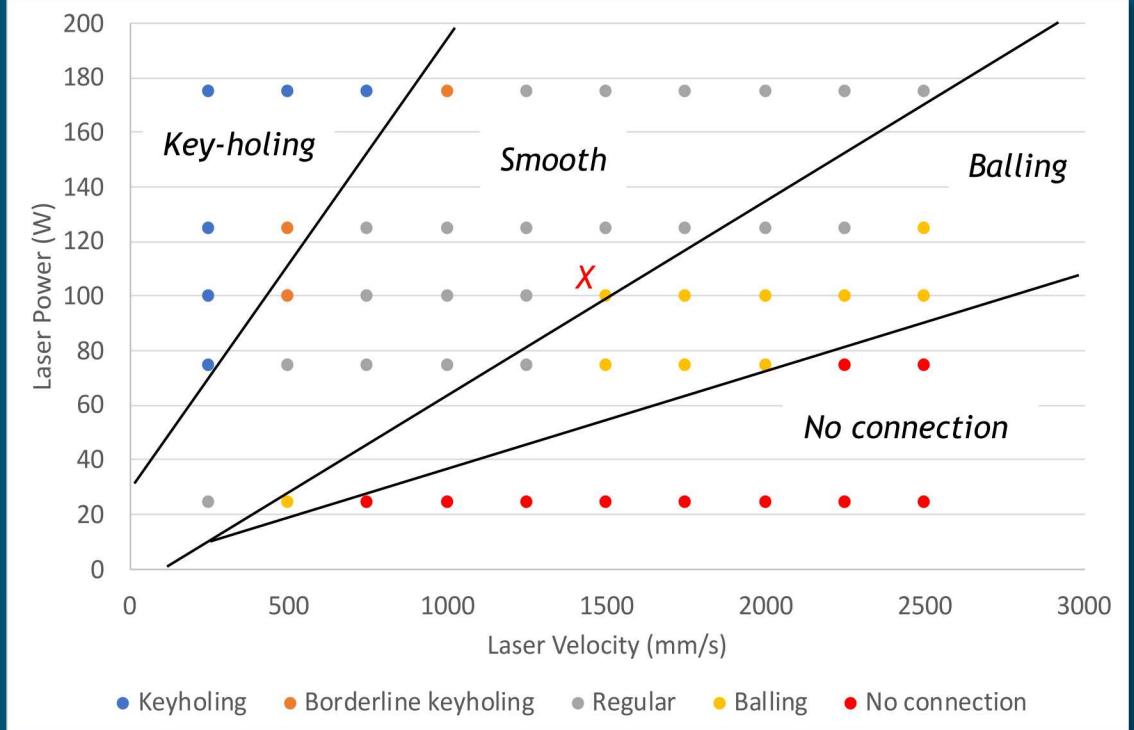
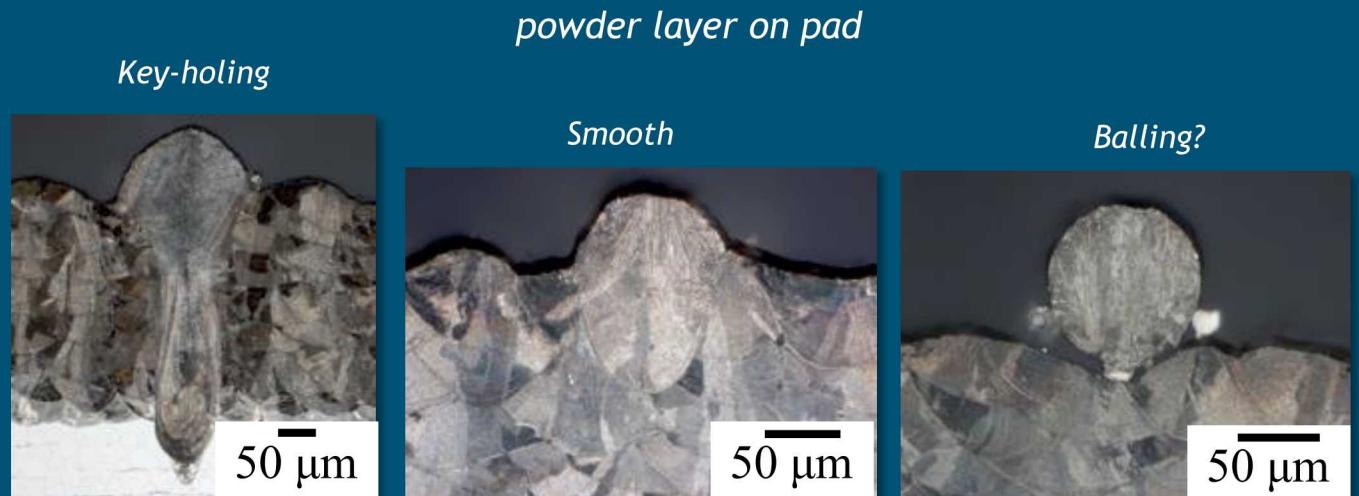
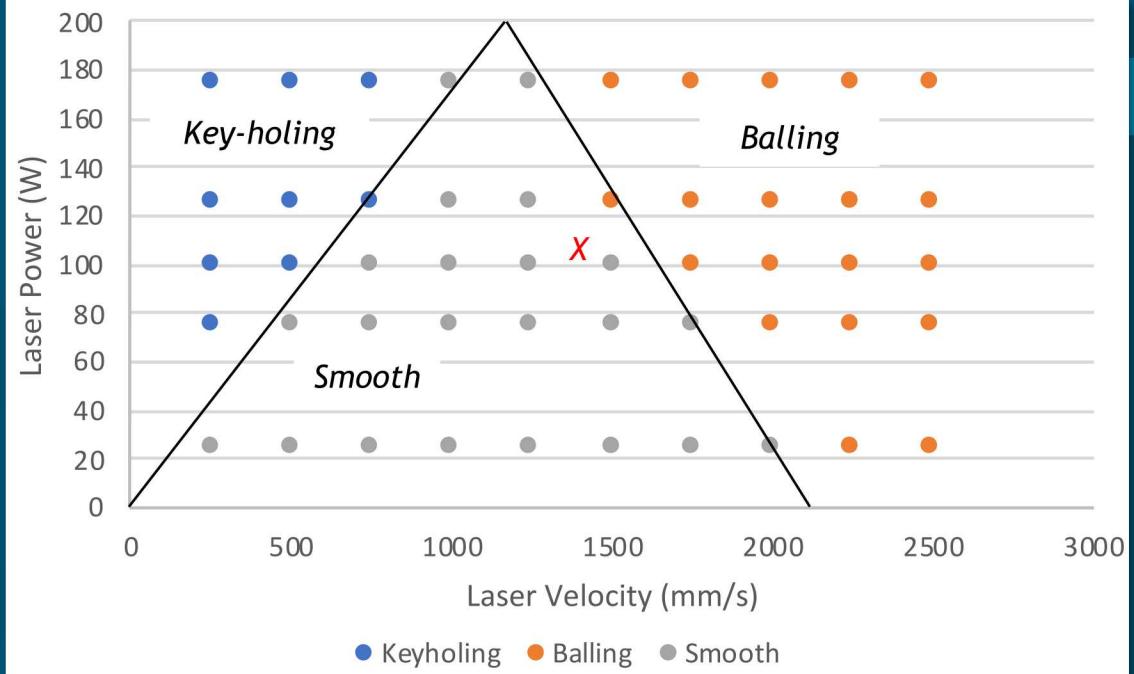
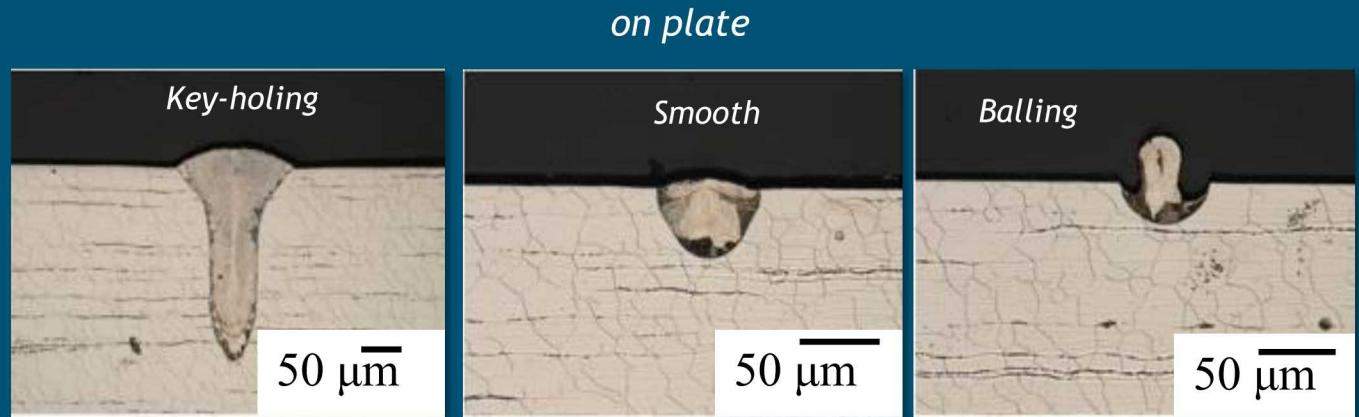


100 W, 250 to 2500 mm/s on pad



melt pool on plate, 100W, 1500 mm/s

Line Scans: Plate vs. Powder Layer on Pad



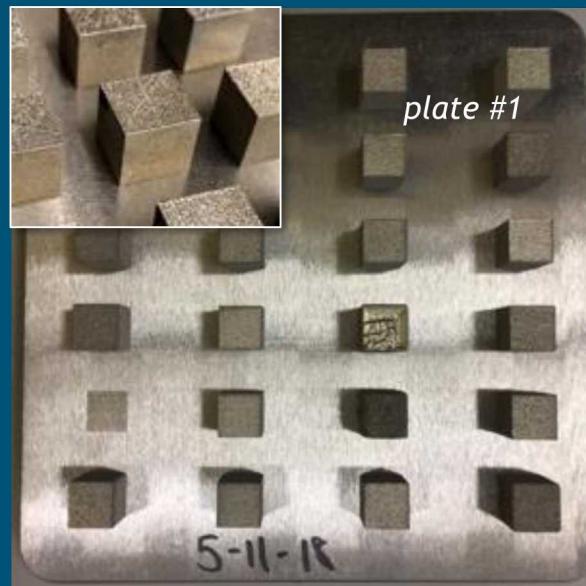
Density Cubes

1cm cubes, 24 per plate

- 10 plates, power, velocity, focus offset, layer thickness, powder size, variation across plate

Bulk material measurements

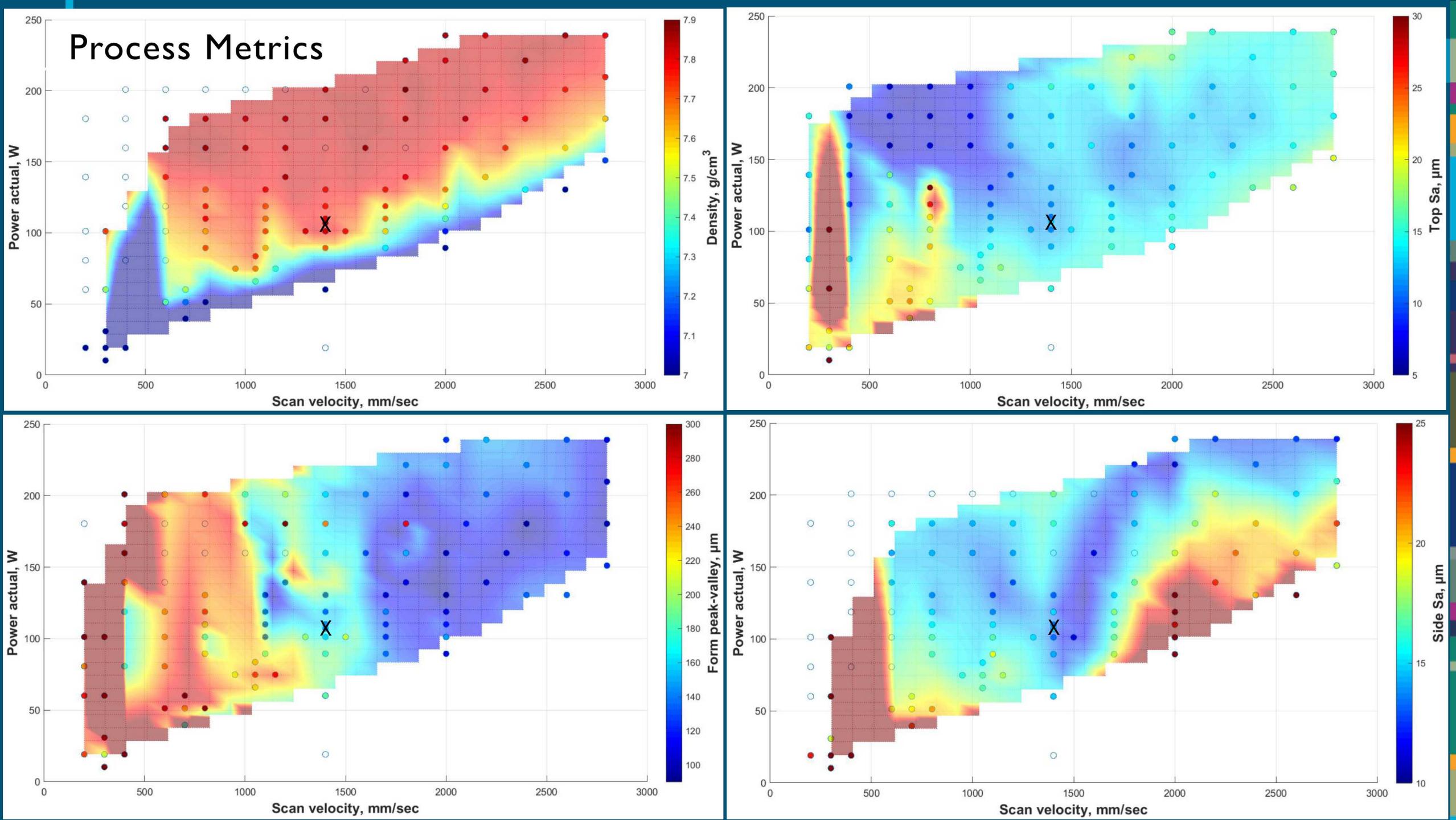
- density (Archimedes)
- top & side surface form & finish
- microstructure: optical, EBSD



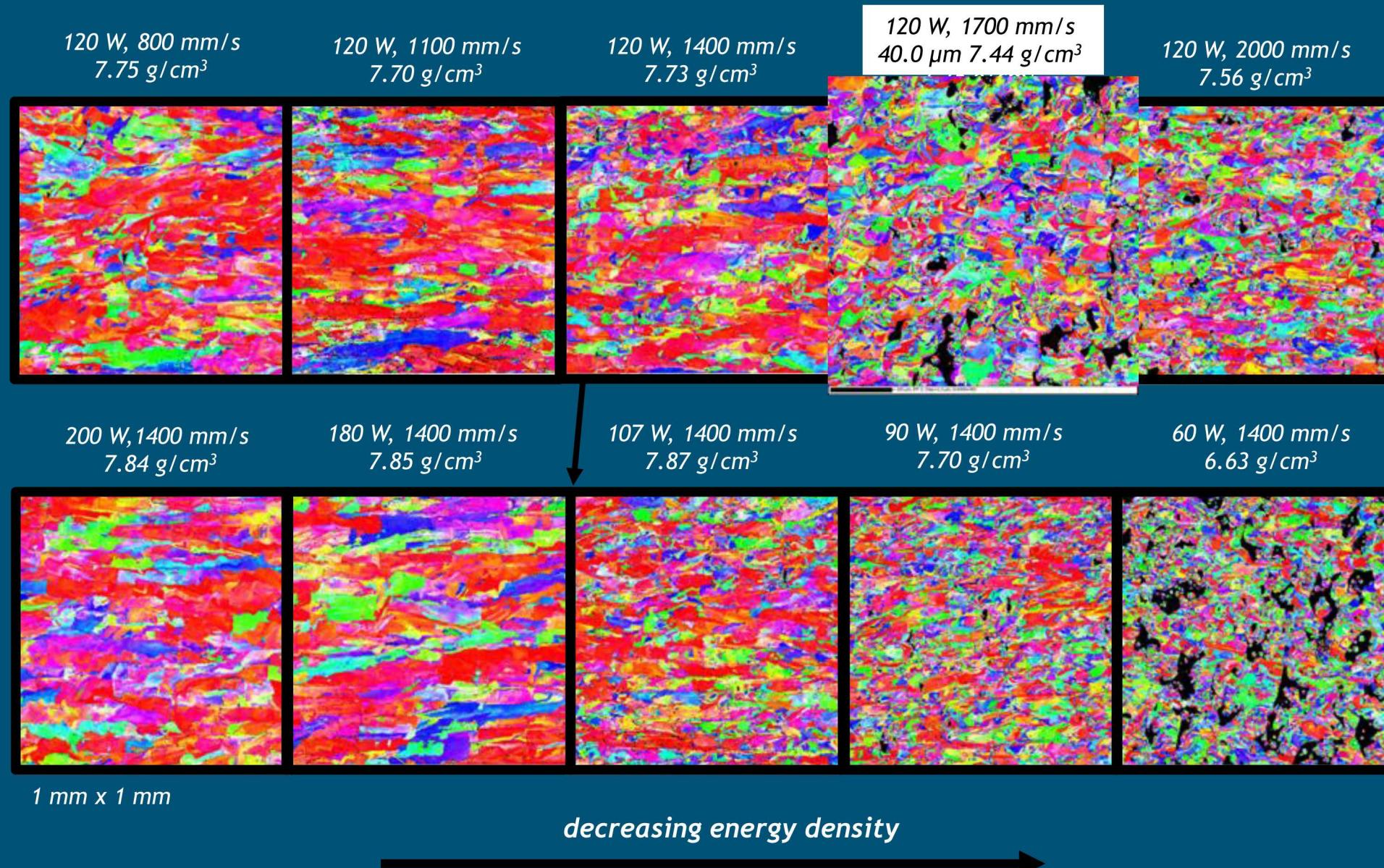
100.2W, 1500mm/sec 209 μm PV form (left), top 13.5 μm Sa roughness (center), side 9.98 μm Sa roughness (right)



60.1W, 300mm/sec 891 μm PV form (left), top 73.8 μm Sa roughness (center), side 38.1 μm Sa roughness (right)



Material Microstructures, 30 μm Layer Thickness



High-Throughput Tensile Properties

1x1x4mm gauge section, 10 or 25 dogbones/array

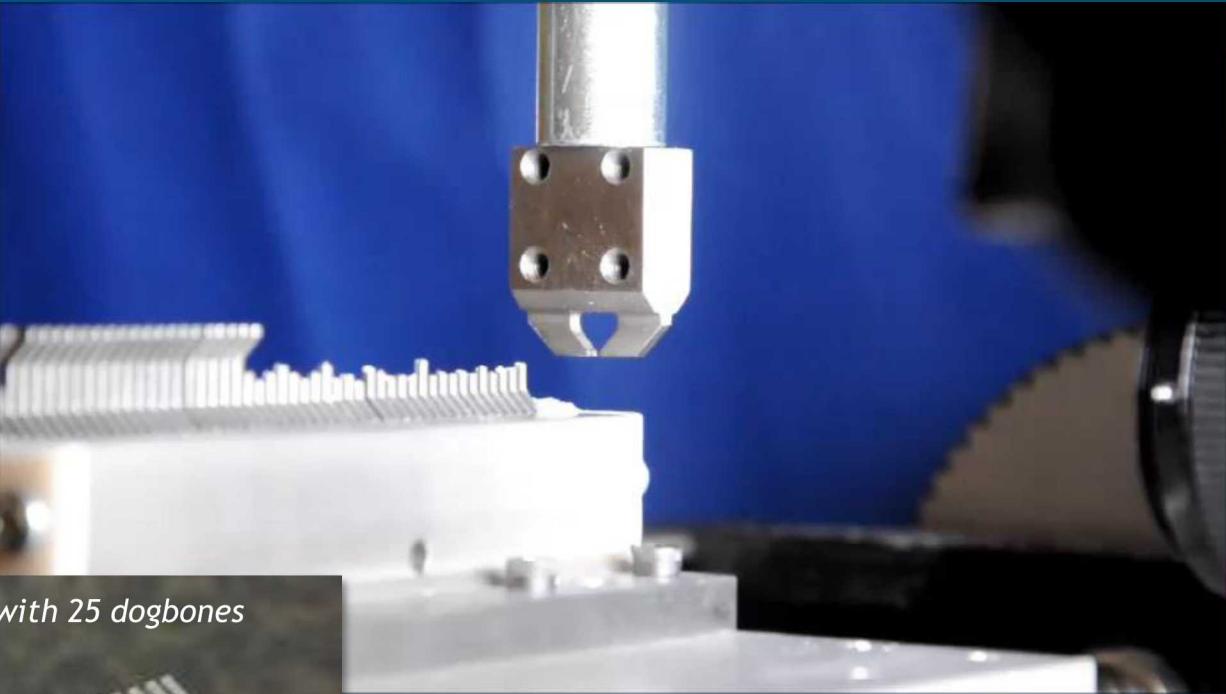
- five build plates
- varied power, velocity, powder diameter
- Gen2 HTT system

>500 dogbones tested

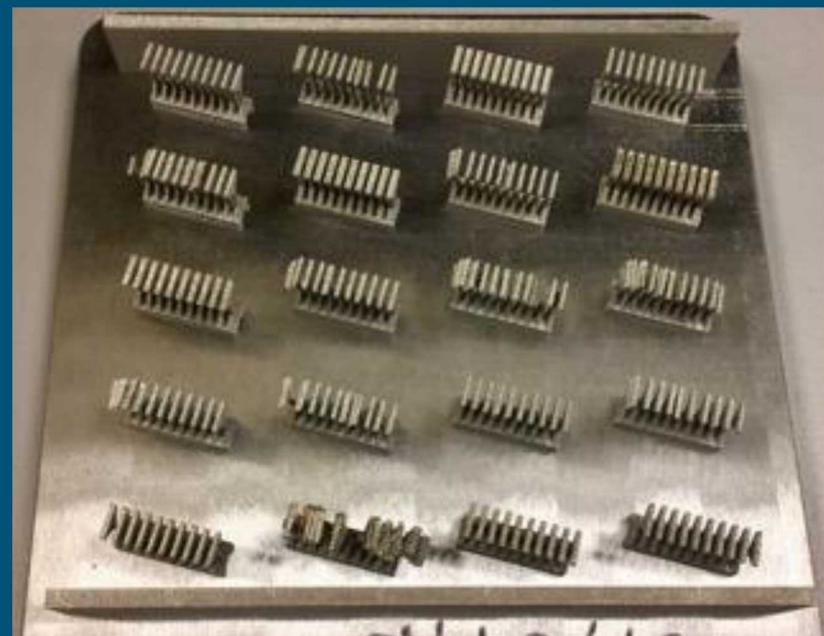
- density (Archimedes)
- surface roughness
- mechanical properties
 - UTS, YS, modulus, ductility

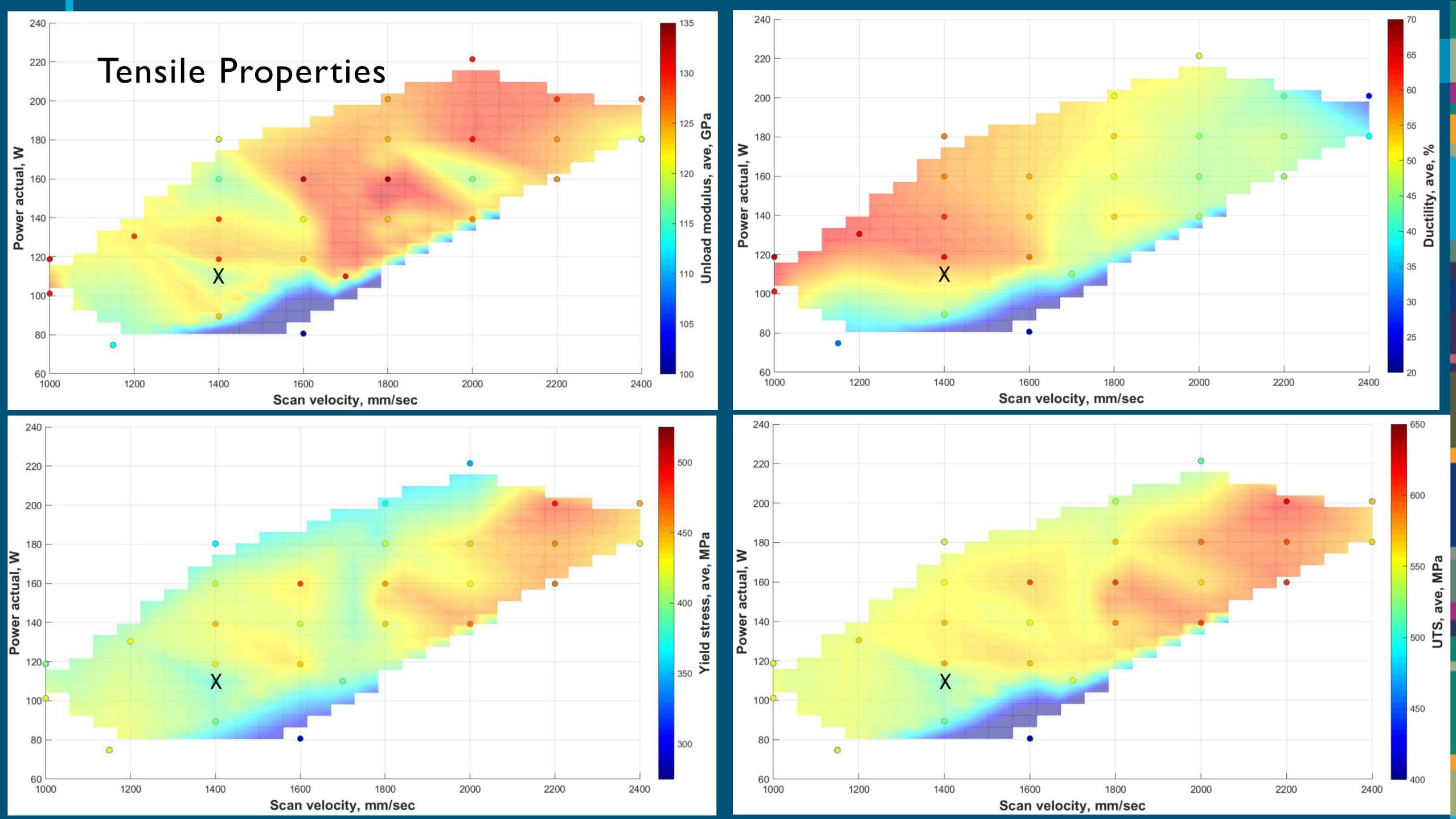


316L SS dogbone array with 25 dogbones



Heckman, Mat. Sci. Eng. A, 2020





Process Understanding & Control Reduced Material Uncertainties



Controlling & logging part, build & powder cycles

- 316L stainless steel
- feedstock pedigree, build environment, process inputs, post-processing, meta data
- print / test artifacts

Heuristic dependent

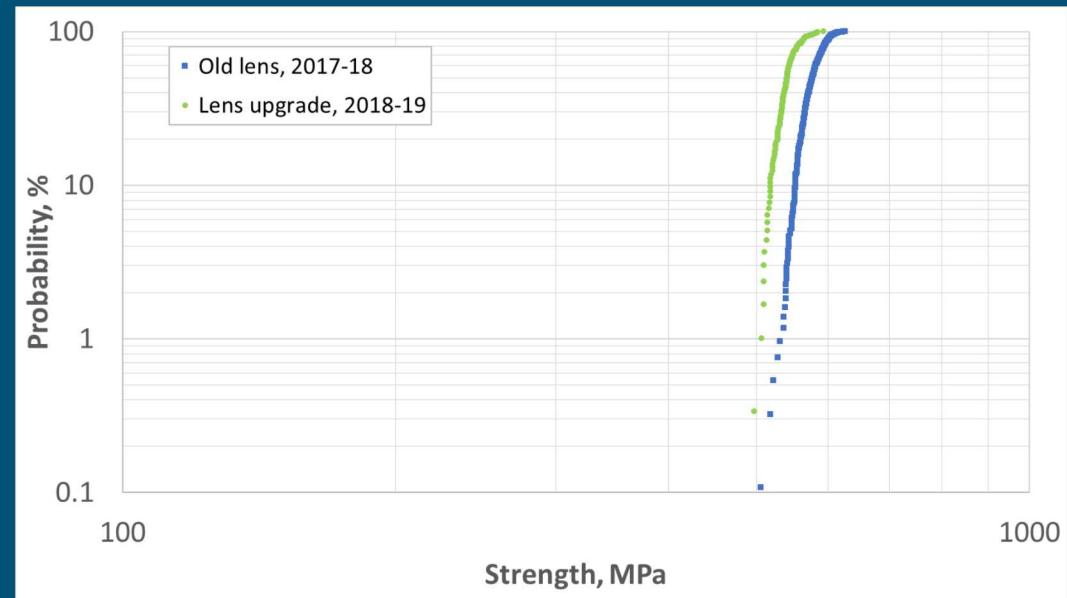
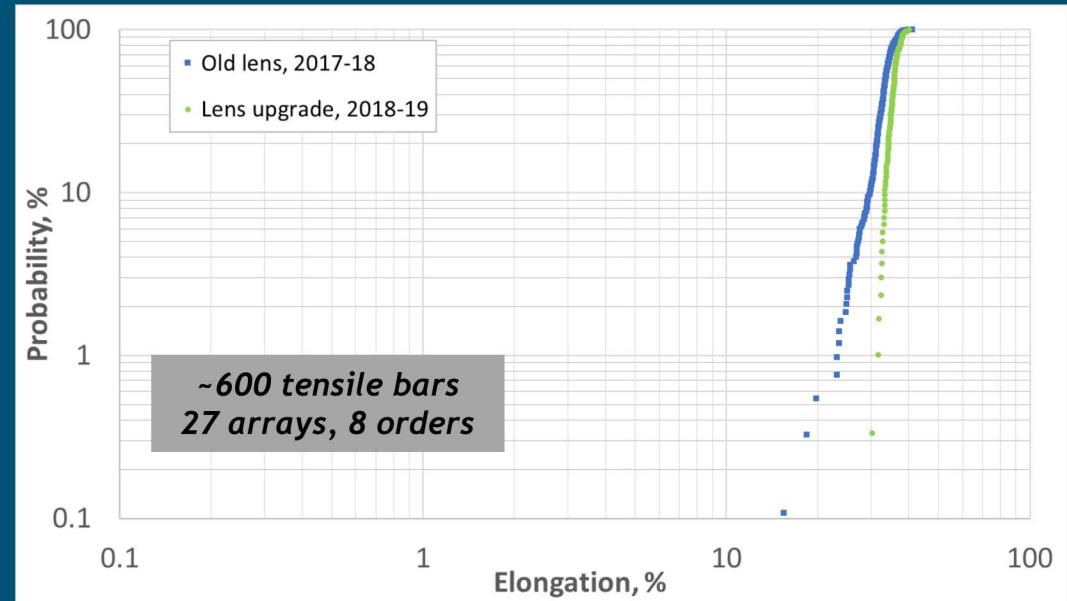
- restrictive
- time consuming
- expensive

Desire accelerated cycles

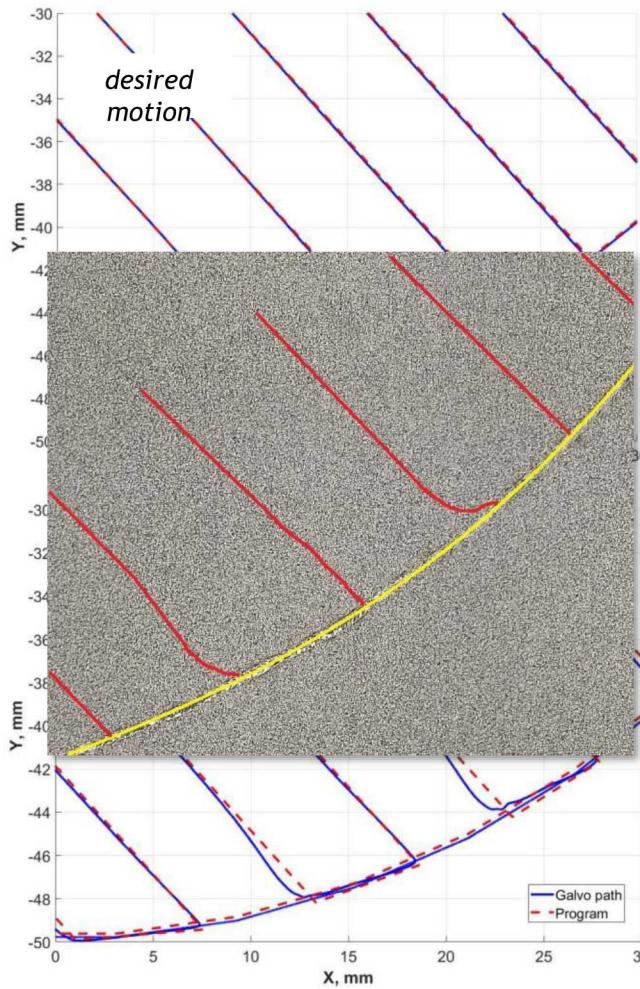
- tolerances & uncertainties



build plate w/process artifacts



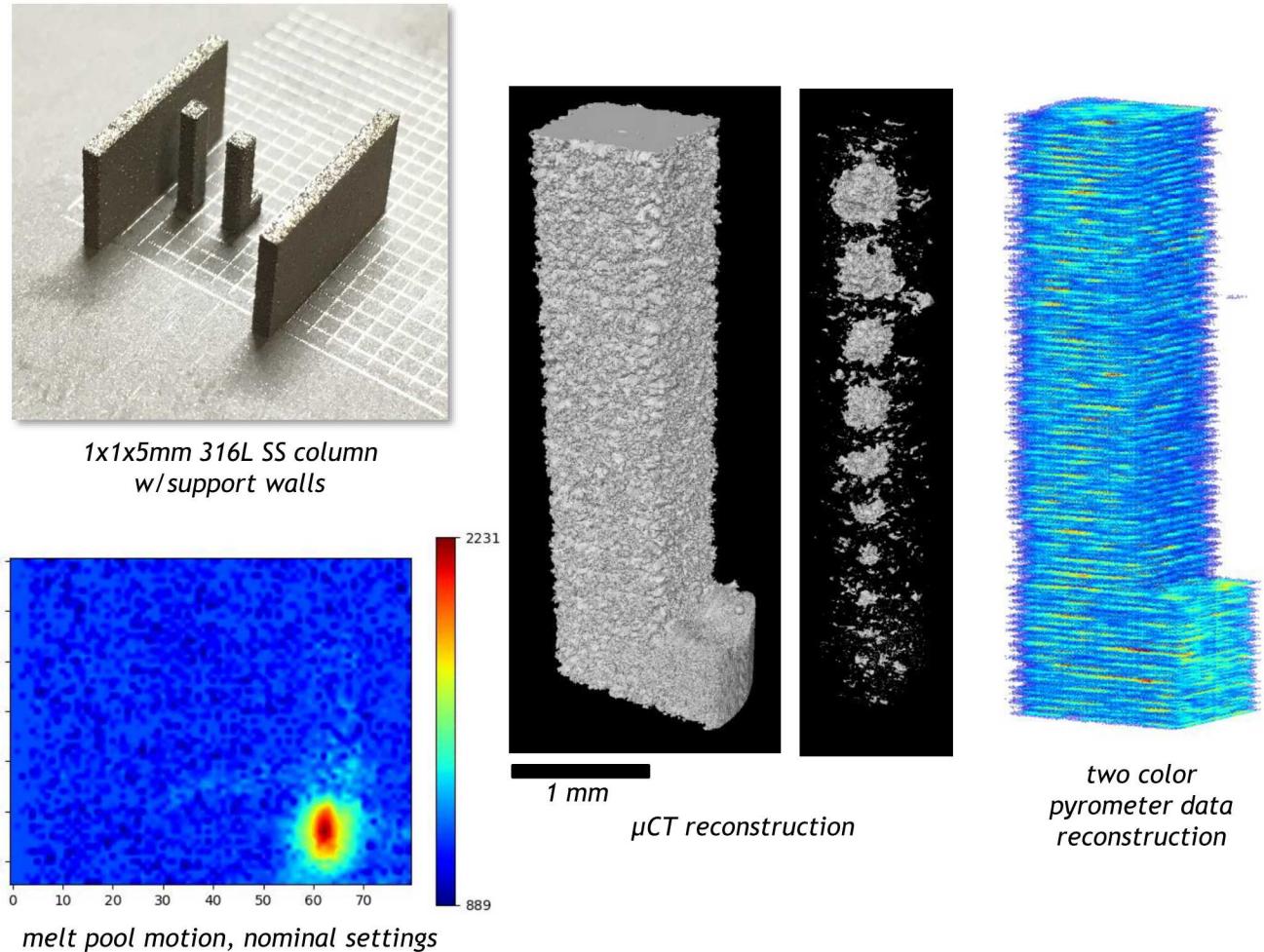
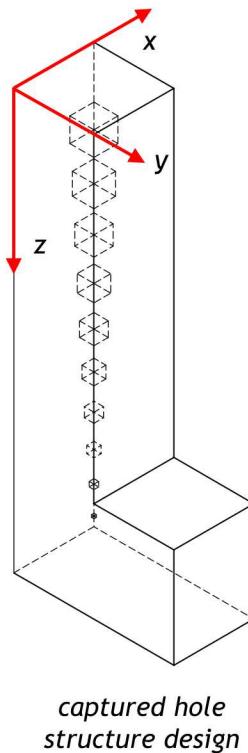
Real-Time Machine Monitoring Reveals Process Perturbations



Archive, Research, Control, Synchronization (ARCS)

- Penn State, 3D Systems collaboration

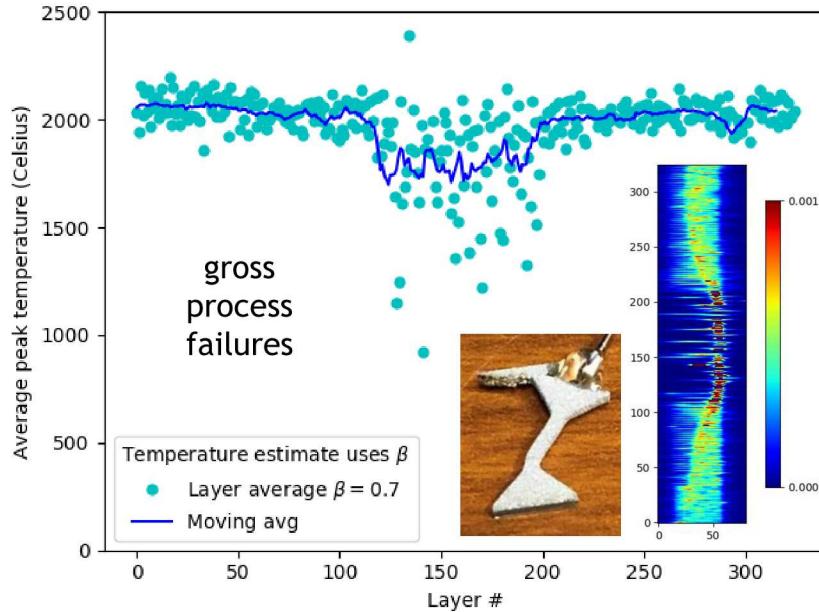
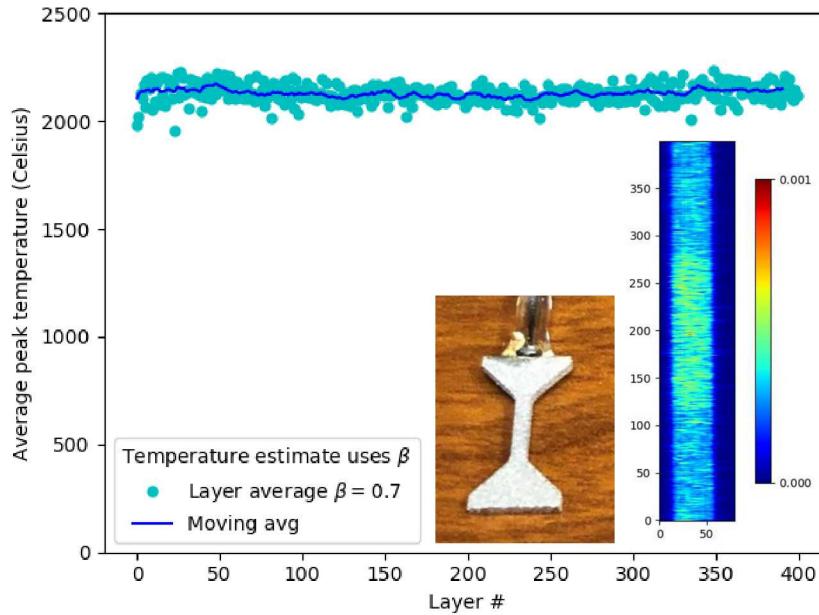
motion @ 100kHz



Exploring multiple sensor signatures

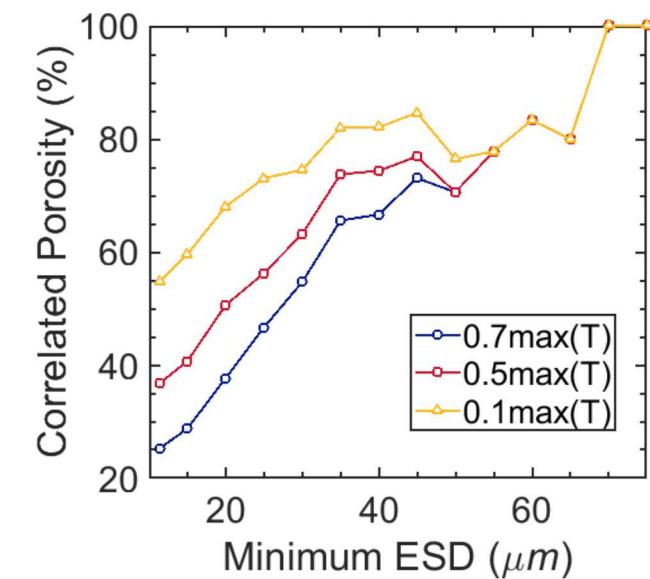
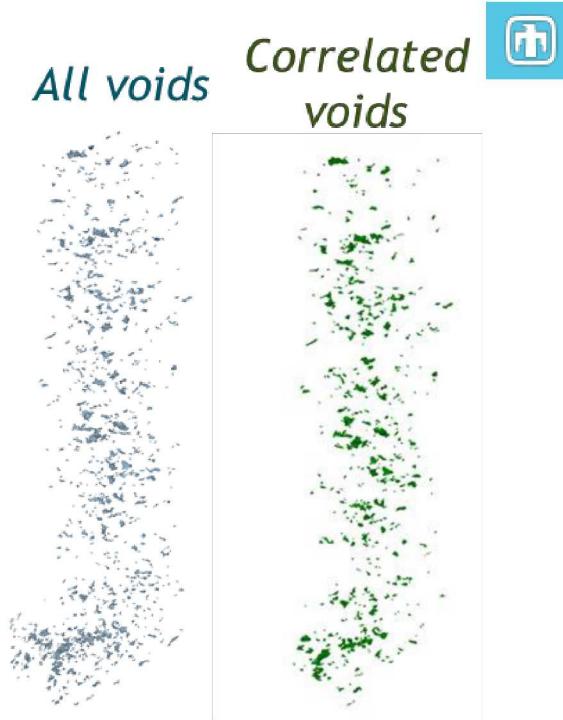
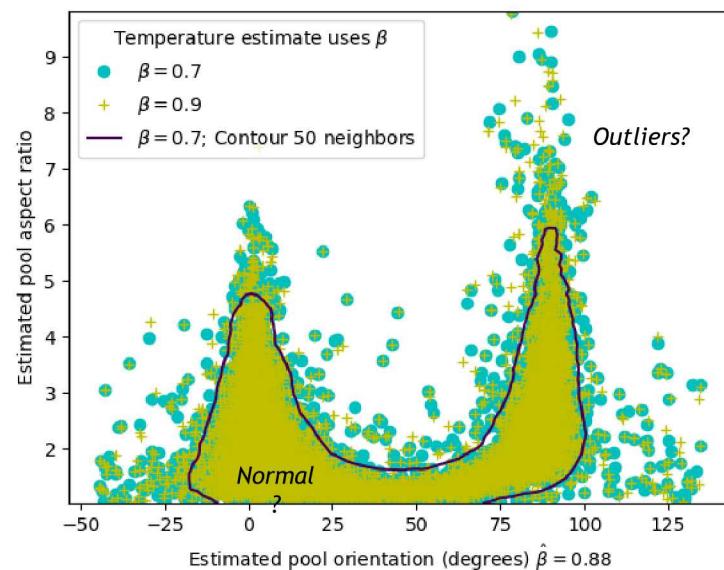
- thermal, optical, acoustic
- two-color pyrometry explored here
- data management & analysis is critical

Correlating Process Signatures w/Material Structures



melt pool outliers
via **machine
learning** correlate
local void porosity

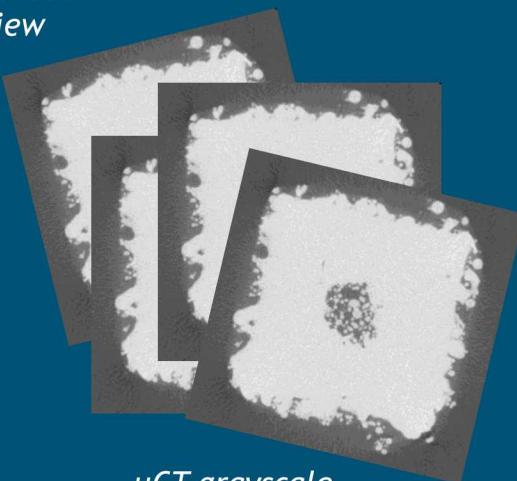
pyrometer data, YZ section (left) & XZ
section (right)



Material Reconstruction



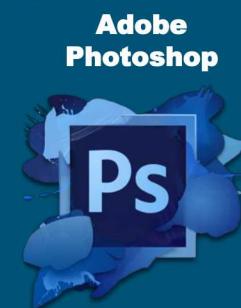
part side view



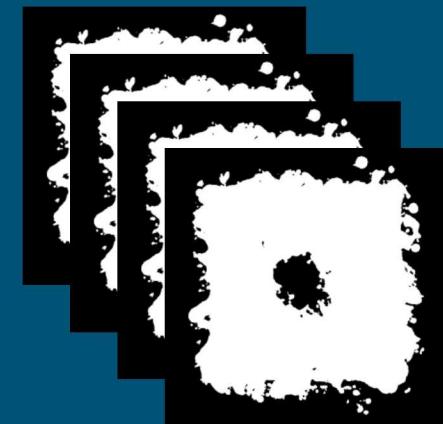
µCT grayscale image stack



micro-computed tomography



batch processing
16bit > 8bit conversion
lossless filetype conversion



aligned & segmented bitmap stack
white = material, black = air/void



alignment &
registration
cropping
grayscale matching
autoleveling
image filtering
thresholding

part reconstruction



Interactive Data Language



reconstruction
quantification



Interactive Data Language

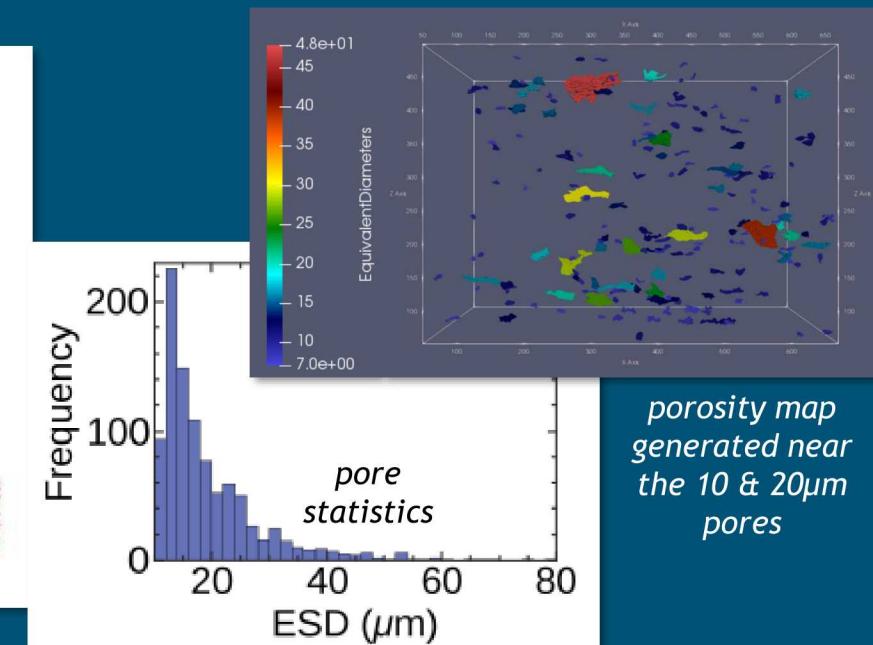


Interactive Data Language



Parallel Visualization Application

image processing
visualization



Data Management



Sharepoint site

- 5Gb storage
- project log – customer, project / program, part CAD files / drawings, requirements, schedule
- part & process log file – build identifier, powder, reuse cycle, process settings, process file name, notes
- as-printed build pictures

Machine computers (lab)

- machine & process build files
- process data – ARCS, in-situ sensor data (thermal, optical, acoustic)
- feedstock data – Aspex SEM image analysis, EDS composition

\\\CEE\Projects\AM Defects\BuildData share folder

- >100Gb storage
- feedstock – powder data, rheology, specification, etc.
- part – CAD file, drawing, specs, build pictures
- process – build file, in-situ process data
- property – tensile, density, Charpy, hardness, corrosion, etc.
- characterization – metallography (optical, SEM, TEM, EBSD), metrology (surface finish, geometry), computed tomography, functional testing

GRANTA

- feedstock, machine, process, material data

project log

part & process log

raw process & part data

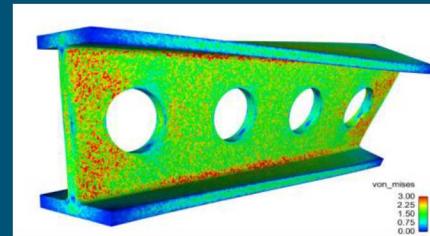
GRANTA

Qualification Tomorrow: Born Qualified

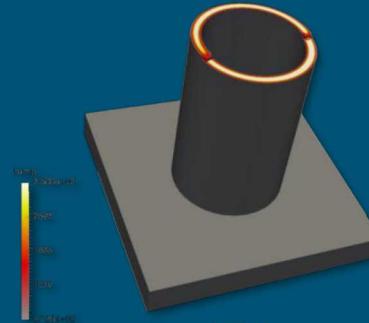


“Changing the Engineering Design & Qualification Paradigm”

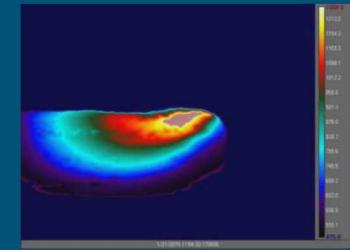
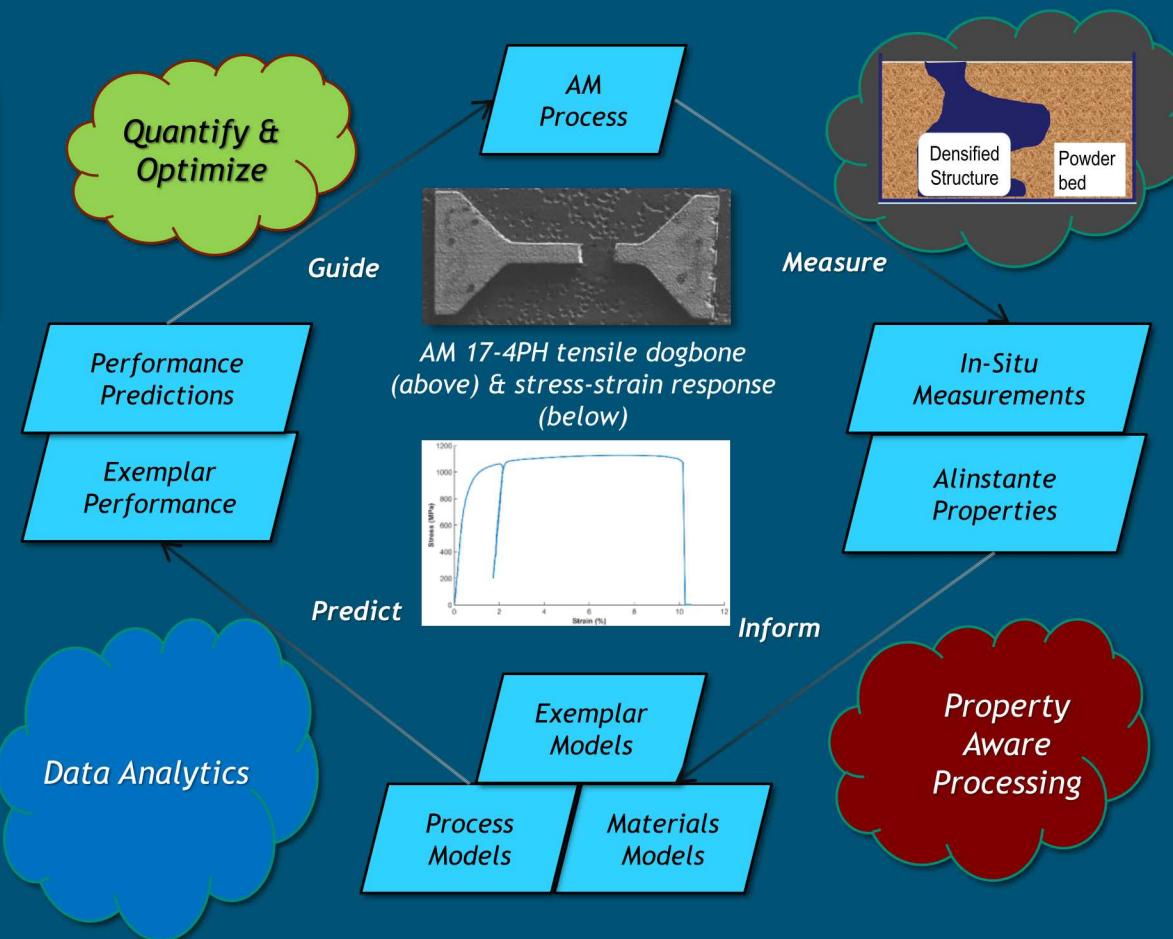
- leverage AM, in-process metrology & HPC to revolutionize product realization
- accelerating design to production



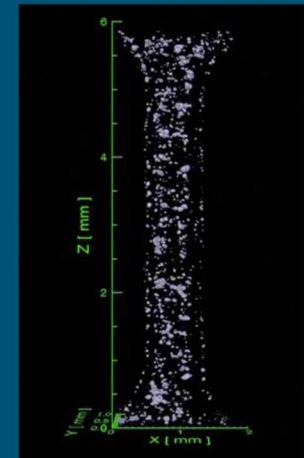
material / part performance simulation



process simulation



thermal history during bi-directional metal deposition



17-4PH dogbone porosity

Models Bridging Length Scales



LAMMPS

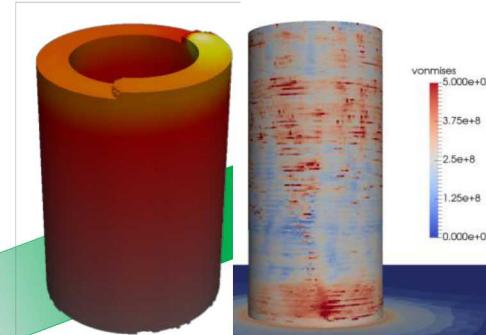
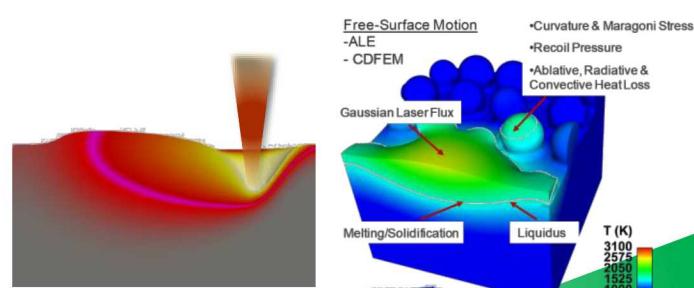
ARIA
ADAGIO
SPPARKS

Build Scale Thermal + Mechanics
K. Johnson, K. Ford, L. Beghini, M. Stender & J. Bishop



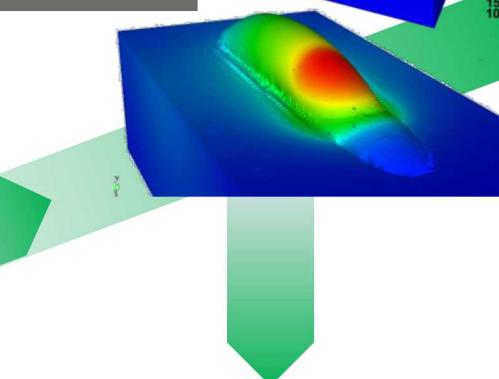
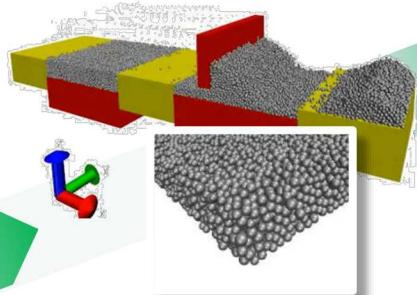
Solidification Scale Thermal

M. Martinez, B. Trembacki, D. Moser

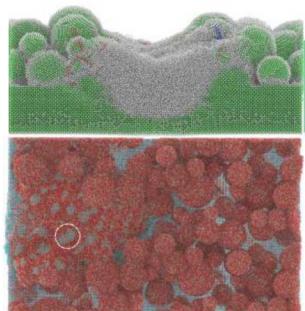


Powder Spreading

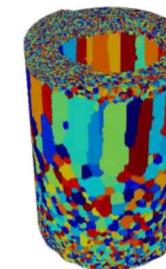
D. Bolintineanu



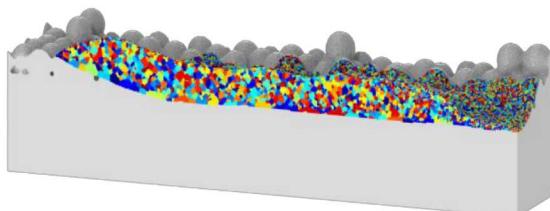
Powder Behavior
M. Wilson



Build Scale Microstructure
T. Rodgers, J. Madison



Mesoscale Texture/Solid Mechanics
T. Rodgers, J. Brown, K. Ford



10^{-6}

10^{-3}

1

Length Scale (m)

Questions?

Bradley Jared, PhD



“Complexity is free” has been an oft-cited refrain during the recent renaissance of manufacturing, typically referencing additive manufacturing (AM). To be sure, however, advanced manufacturing techniques make complexity available as new materials, processes and part topologies are realized enabling performance gains inaccessible through traditional means. Such gains will be discussed in the context of laser-powder bed fusion (L-PBF). The development of physics-based engineered controls and automation methodologies are preferred to satisfy requirements by optimizing process performance, by minimizing output uncertainties, and by increasing product throughput.

Research on L-PBF will be presented to address these process complexities and performance uncertainties. Initial material behavior mimicked an open loop state as properties varied widely and proved difficult to anticipate. Work has since quantified important process-structure-property relationships for powder feedstocks, laser settings and machine components. Machine improvements and metrology procedures have also been developed, producing an increase in process consistency and material confidence. Continued research is examining in-situ melt pool dynamics and exploring advanced data analytic techniques (ex. machine learning) to correlate melt pool outliers to process defects.