

Additive Manufacturing Process Modeling – Progress Towards Science-Based Certification

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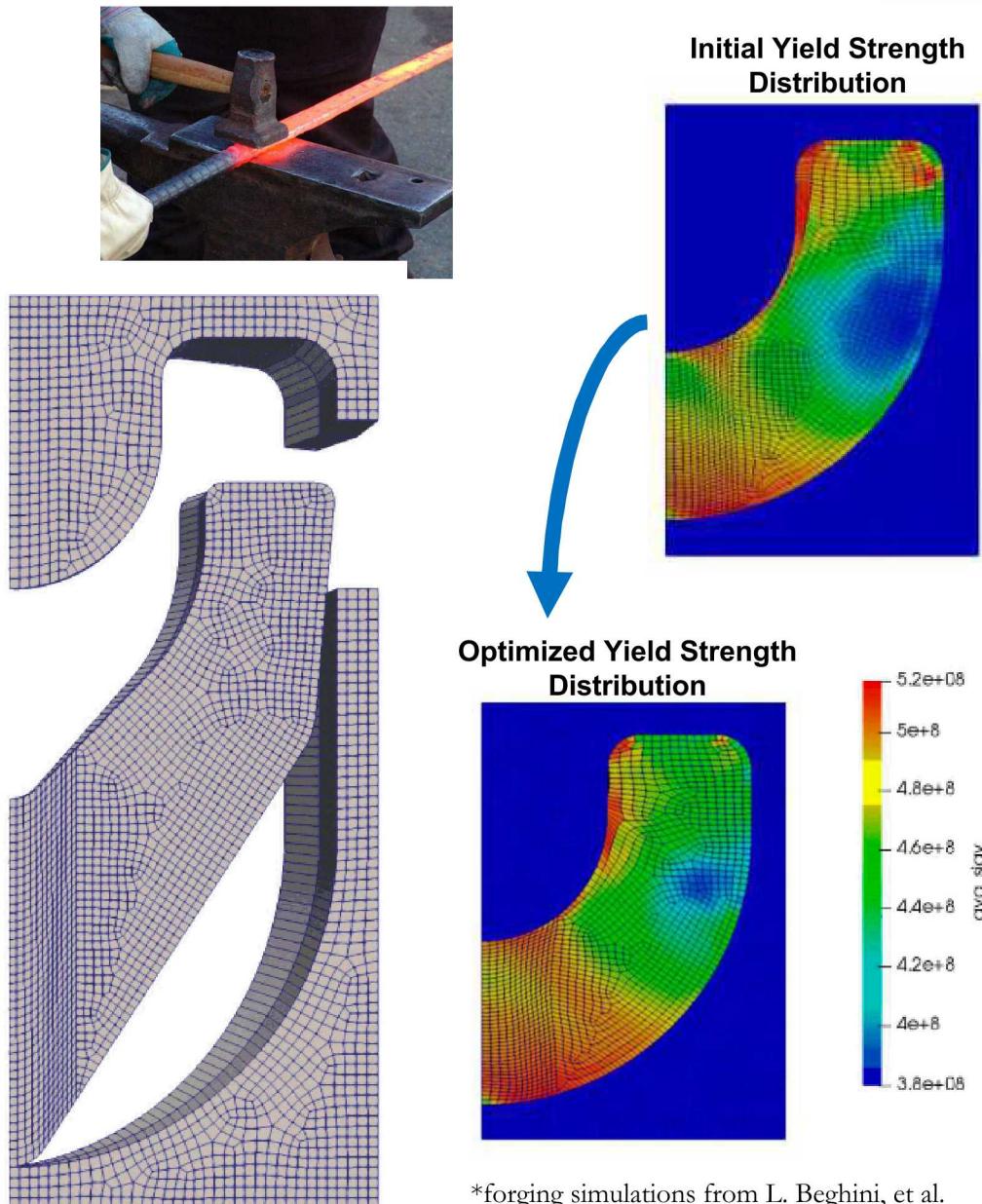
Sandia – UC Davis Research Partnership Symposium

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Additive manufacturing – integrated computational materials engineering to accelerate development

- Historically, traditional manufacturing methods have been developed over long periods of time
 - Large empirical data sets, trial and error, etc.
- At Sandia, computational engineering has been employed to optimize traditional manufacturing processes (e.g. forgings)
- Additive manufacturing presents many advantages, yet requires an accelerated development timeline
 - Computational simulation should be applied to accelerate development and enhance understanding for additive manufacturing to improve outcomes
 - Certification of properties and performance for AM parts requires additional research and development activities

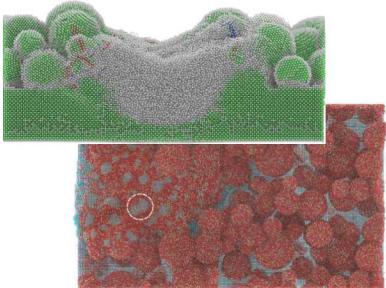


A variety of modeling tools for additive manufacturing have been developed by Sandia National Laboratories

Simulation Codes:
LAMMPS, SPPARKS,
SIERRA/Aria, SIERRA/Adagio

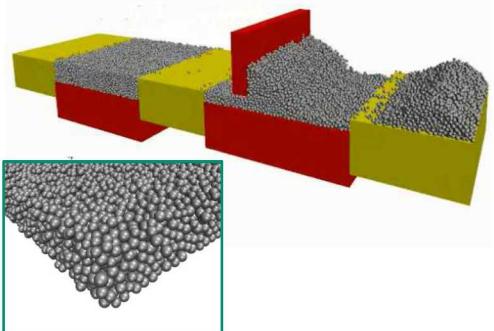
Powder Behavior

Mark Wilson



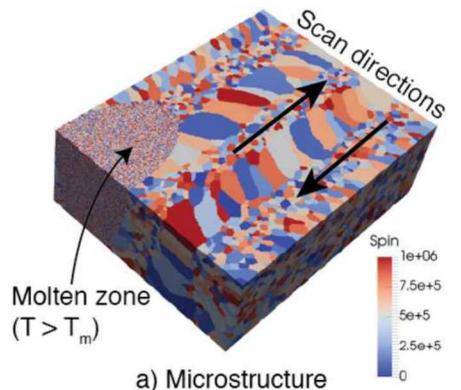
Powder Spreading

Dan Bolintineanu



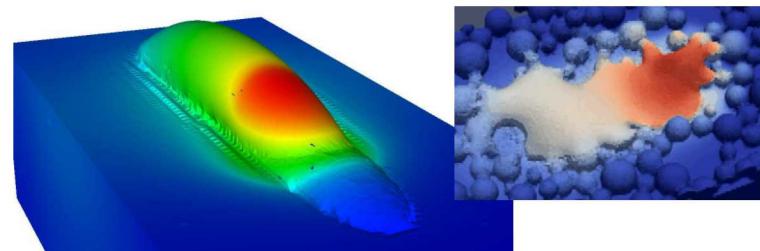
Mesoscale Texture/Solid Mechanics/CX

Judy Brown, Theron Rodgers and Kurtis Ford



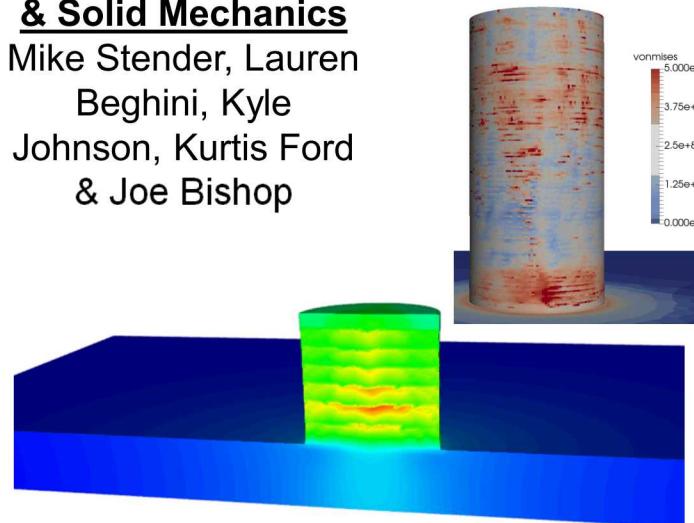
Mesoscale Thermal/Fluid Behavior

Brad Trembacki, Dan Moser & Mario Martinez



Part-Scale Thermal & Solid Mechanics

Mike Stender, Lauren Beghini, Kyle Johnson, Kurtis Ford & Joe Bishop



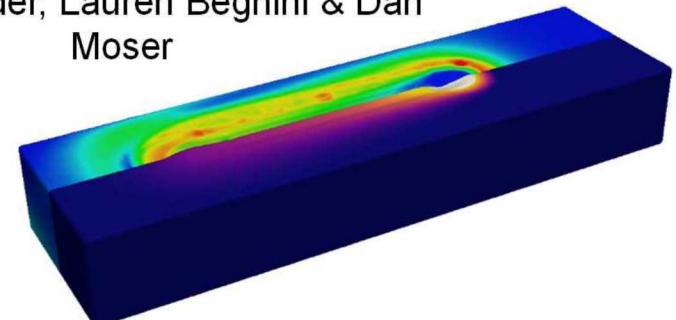
Part-Scale Microstructure

Theron Rodgers



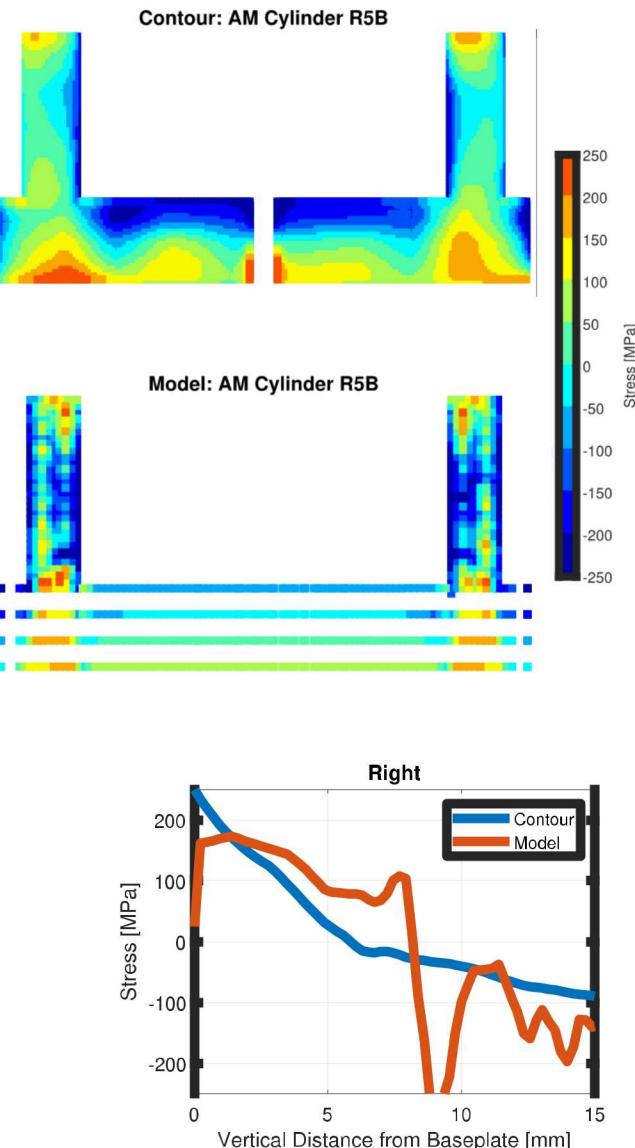
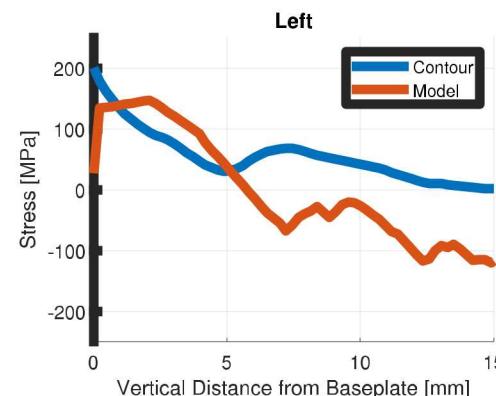
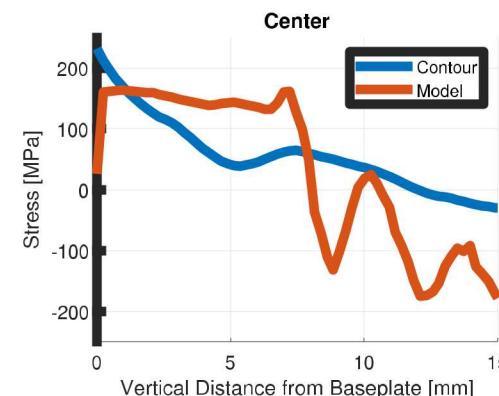
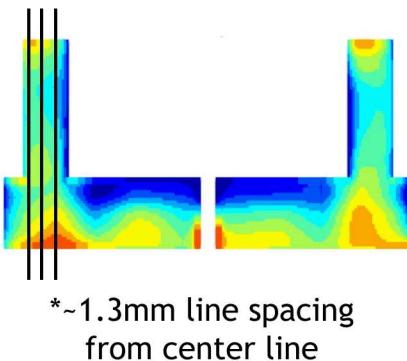
Mesoscale Coupled Solid/Fluid Behavior

Mike Stender, Lauren Beghini & Dan Moser



Work completed in Partnership with UC Davis has provided valuable validation information

- Campus Executive LDRD projects have supported initial process model validation activities in FY19
 - Prof. Mike Hill, Chris D'Elia, Nick Bachus – residual stress measurement in additive manufactured parts
 - Prof. Mark Rashid, Madison Richey – efficient simulation for additive manufacturing
- Measurements from Mike Hill and Chris D'Elia shown here have provided essential residual stress data for model validation



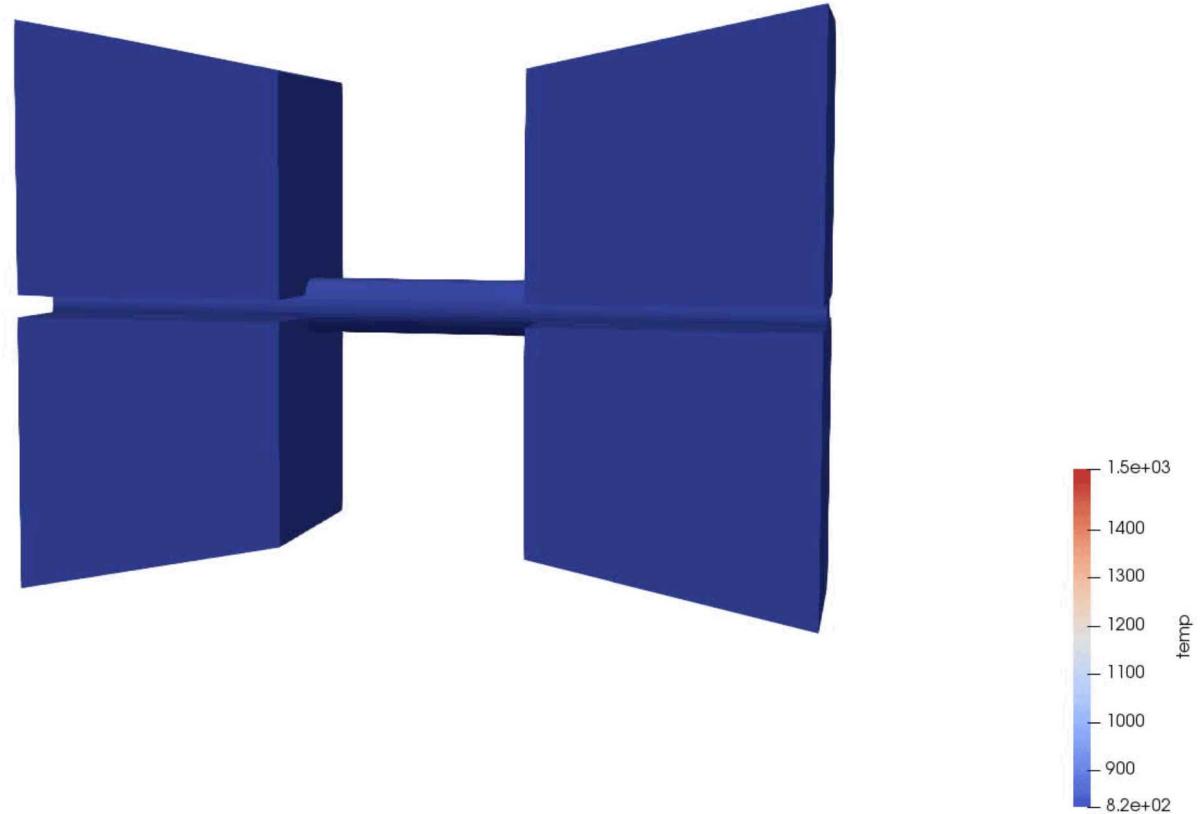
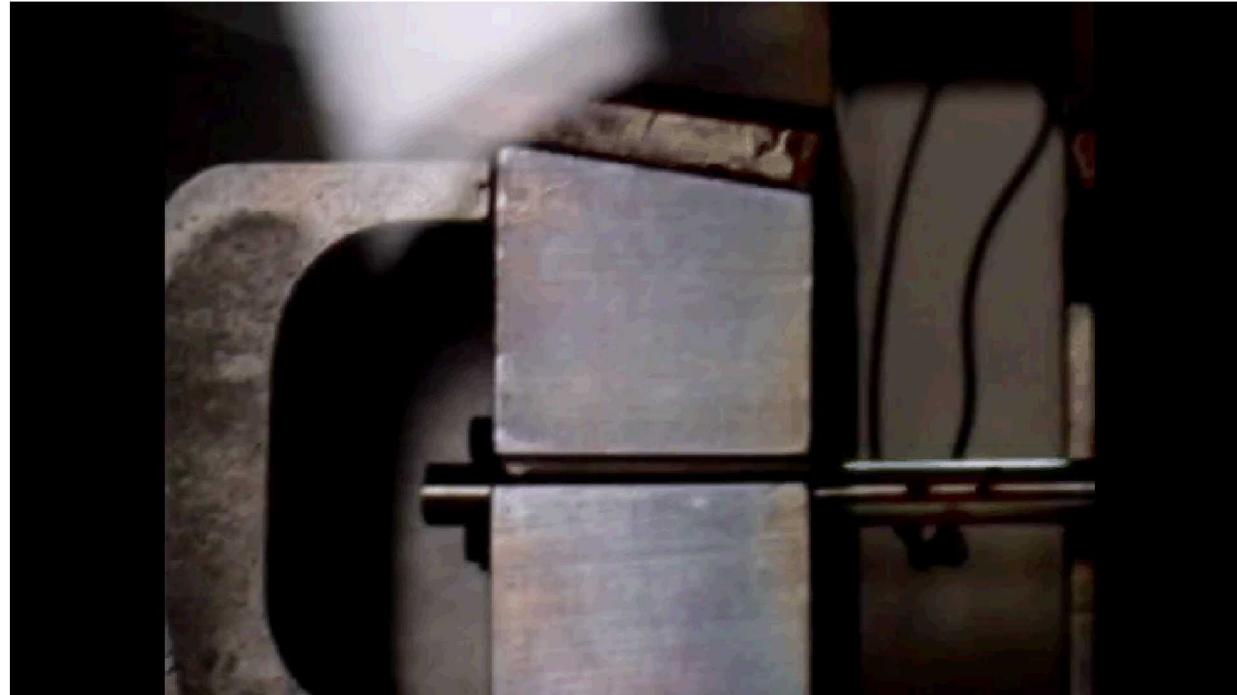
Tractable challenges for additive manufacturing simulation

- Many opportunities exist to improve outcomes for additively manufactured components
- Process improvements and additive manufacturing specific opportunities
 - Laser optimization – scan path, active power control, thermal mechanical history control
 - Residual stress engineering
 - Engineering of material properties to optimize performance
 - Hybrid and functionally graded materials
- Continued validation of initial model predictions is essential
 - It is critical to compare across measurement techniques (e.g. contour method, neutron diffraction, slitting) whenever possible
 - Similarly, different modeling strategies (e.g. solid-fluid coupled, part-scale) should be compared and assessed against each other
- Perfect validation is unlikely, understanding differences and quantifying uncertainties in models and experiments is critical

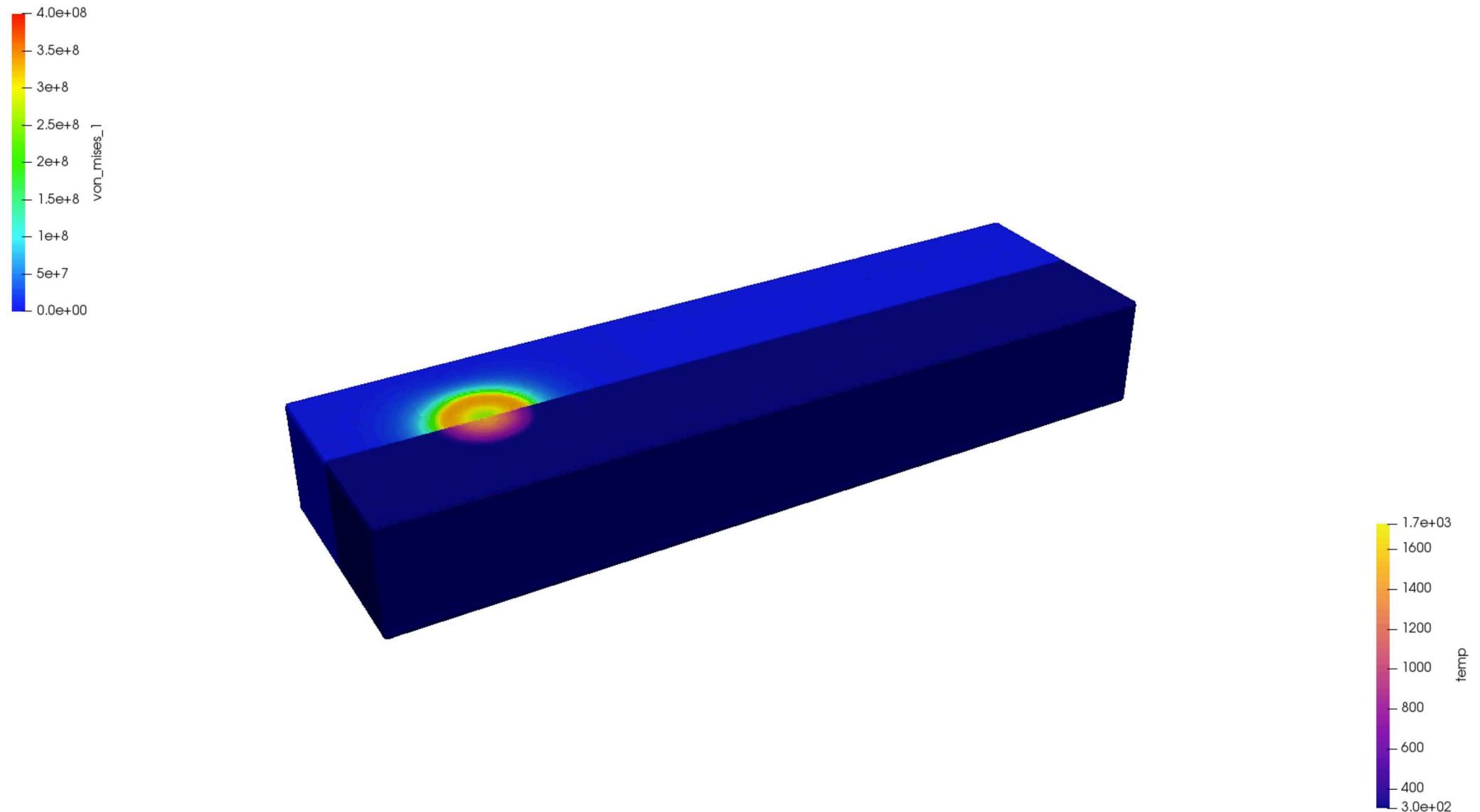


Constitutive model calibration is leveraging Gleeble tests

- Gleeble tests are underway to calibrate high temperature material parameters



High-fidelity model shows improvements in residuals stress calculations



Constitutive Model 304L Stainless Steel

- Elastoviscoplastic temperature dependent material model calibrated for 304 L (BCJ mem)
 - Calibrated for room temperature to forging temperatures (< 1200 K)
 - Continuing work into higher temperature calibration up to near melt (~1700 K)
 - Temperature dependent thermal and mechanical properties

Flow rule

$$\dot{\epsilon}_p = f_1 e^{-f_2/\theta} \sinh^{n_1 + \frac{n_2}{\theta}} \left(\frac{\sigma}{\kappa + Y(\theta)} - 1 \right)$$

Temperature dependent yield stress

$$Y(\theta) = \frac{1}{2} \frac{Y_0}{Y_4 + e^{(-\frac{Y_1}{\theta})}} [1 + \tanh\{Y_2(Y_3 - \theta)\}]$$

If $T > T_{melt}$ then,

$$T_{ij} = -p\delta_{ij} + \mu_{melt} \left(\frac{\partial v_i}{\partial x_j} + \frac{\partial v_j}{\partial x_i} \right)$$

