

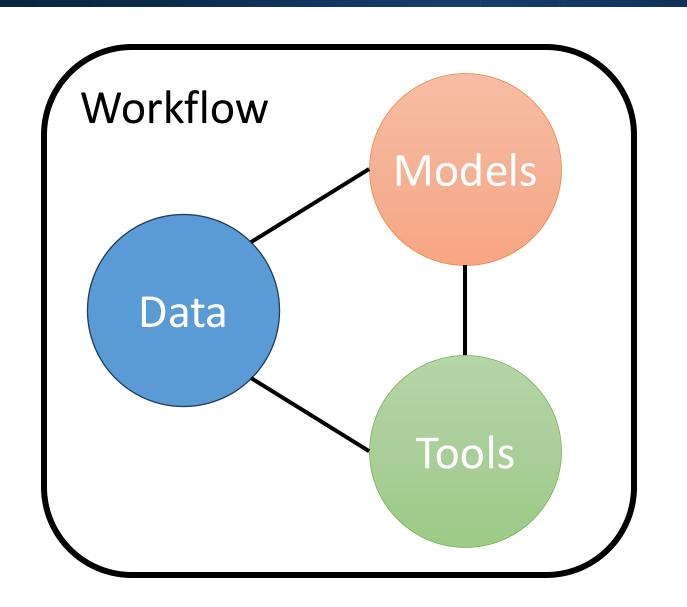
# Toward the Verification, Validation, and Uncertainty Quantification Strategy for Digital Twins

Shengyen Li Engineering Laboratory, NIST

Integrated Computational Materials Engineering for Material Design

## Trouble with Integration





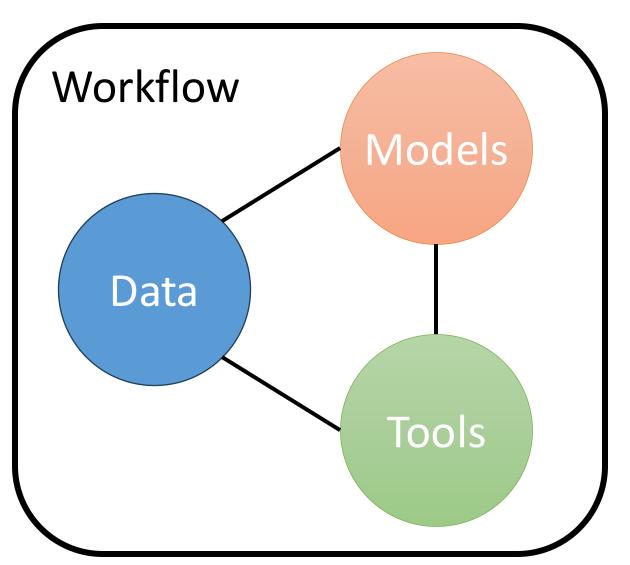
**Engineering Application** 

#### Data sources

- https://www.nist.gov/programsprojects/configurable-data-curation-system-cdcs
- https://ambench2022.nist.gov
- https://ambench.nist.gov

# Trouble with Modeling





#### **Sample Models**

Physics-based model

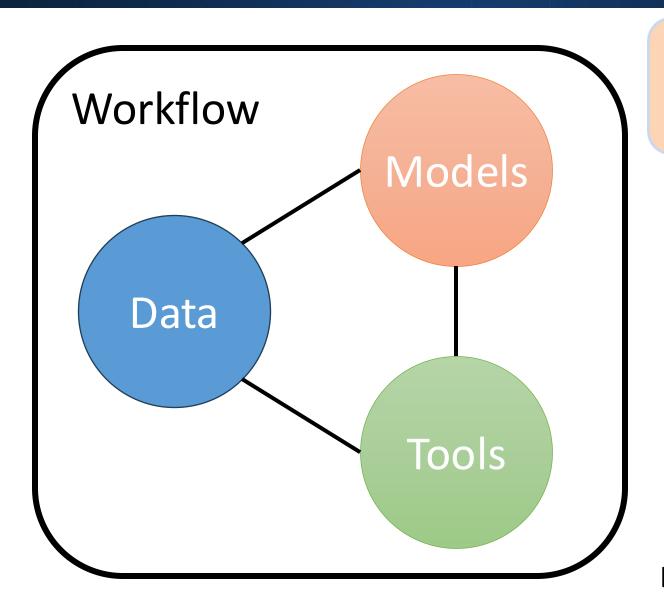
$$\dot{N}(t) = Z\beta N_0 \exp\left(\frac{-\Delta G^*}{K_B T}\right) \exp\left(\frac{-\tau}{t}\right)$$

Polynomial regression  $y = f(x_1, x_2, x_3 \dots)$ 

Neural network 
$$y = \cdots f_2(f_1(x_1, x_2, f_2(x_3) ...))$$

### Trouble with Workflow





#### Homogeneous NON-Steady-State Nucleation

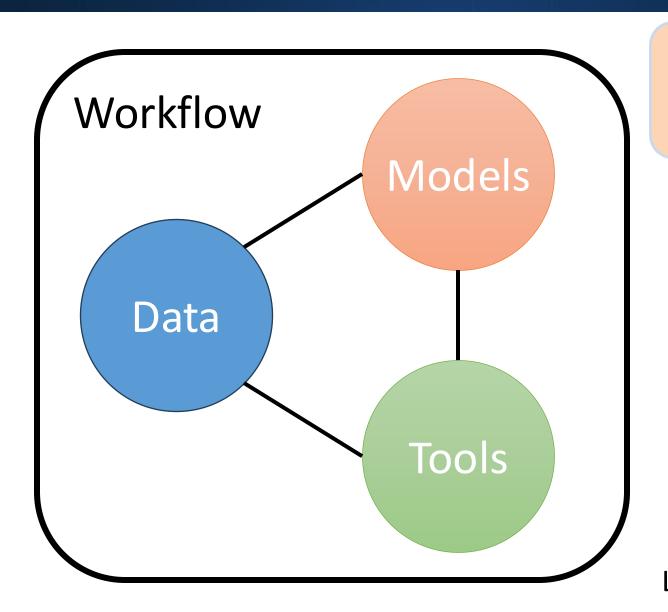
Russell, 1978; Wagner et al., 2001

**Growth/Coarsening**  $\dot{N}(t)$ Rougier et al., 2013; Perez et al., 2008  $C_i^{\gamma} - \bar{C}_i^{\gamma}$  $\sigma_{YS} = \sigma_0 + \sigma_{SS} + \sigma_{H-P} + \sqrt{\sigma_{\rho,0}^2 + \sigma_p^2}$  $\mathbf{F} = \left[ \left( 6 \bar{R}^{\gamma'} \mathbf{F}_{ABB} V^{\gamma'} \right)^{0.5} \right]$ **Elastic Deformation: PyMKS** 0.5 752

Li2017Computational

### Trouble with Workflow Trust





#### Homogeneous NON-Steady-State Nucleation

Russell, 1978; Wagner et al., 2001

**Growth/Coarsening**  $\dot{N}(t)$ Rougier et al., 2013; Perez et al., 2008  $C_i^{\gamma} - \bar{C}_i^{\gamma}$  $\sigma_{YS} = \sigma_0 + \sigma_{SS} + \sigma_{H-P} + \sqrt{\sigma_{\rho,0}^2 + \sigma_p^2}$  $\mathbf{F} = \left[ \left( 6 \bar{R}^{\gamma'} \mathbf{F}_{ABB} V^{\gamma'} \right)^{0.5} \right]$ **Elastic Deformation: PyMKS** 0.5

Li2017Computational

### Trustworthiness of Workflow



- Trustworthiness of simulations, hardware, integration and beyond
- A checklist to assess the trustworthiness of simulations in the 4 phases of the development lifecycle

Definition	Development	Deployment	Maintenance
Context of Use	Problem Space	Hardware Integration	Hardware maintenance
Objectives	Measurable Phenomena	Data Integration	Recalibration Frequency
Tolerance	Data Management	Problem Spaces Comparison	Backward Compatibility
Resources	Data Variability	Uncertainty Propagation	
Constraints	Uncertainty Quantification	Workflow Refinement & Validation	
	Computational Detail	Scalability	

### Trustworthiness of Simulations

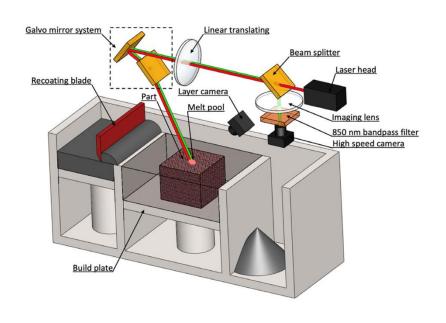


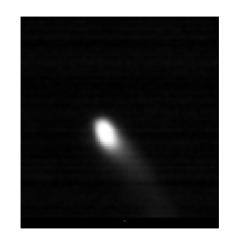
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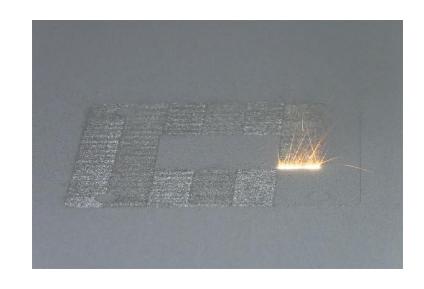
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### Use case for Additive Manufacturing



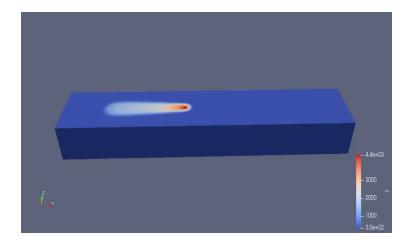






N0001 G01 X57.5000 Y-9.4974 Z-3.4790 F1000 L0 D1.00 N0002 G01 X57.4989 Y-9.5000 L285 N0003 G01 X57.3794 Y-9.5000 L0 N0004 G01 X57.5000 Y-9.2160 L285 N0005 G01 Y-8.9345 L0 N0006 G01 X57.2600 Y-9.5000 L285 N0007 G01 X57.1405 Y-9.5000 L0 N0008 G01 X57.5000 Y-8.6530 L285 G-COCE N0009 G01 Y-8.3715 L0

Model

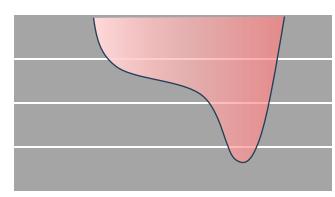


### Challenges



- Laser Power
- Scanning Speed



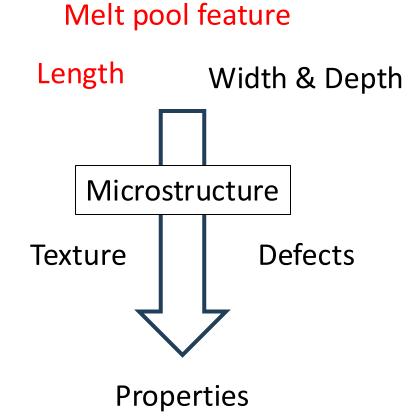


- Location: X, Y, Z
- Laminar Gas Flow



4,770,436 images

- Material
- Layer Thickness
- Subtract Temperature



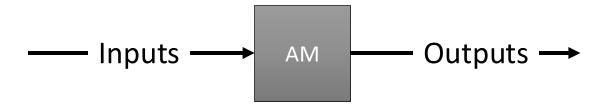
Lane2020Process
Yang2022Investigating

"... there are over 100 of such processing parameters...,"

Oliveira 2020 Processing

# **Problem Space**





#### Command file

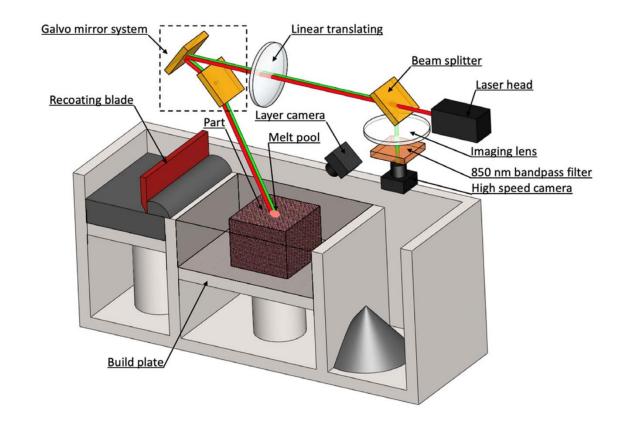
- Laser location
- Laser power
- Time

#### Data acquisition file

- Laser location
- Laser power
- Time

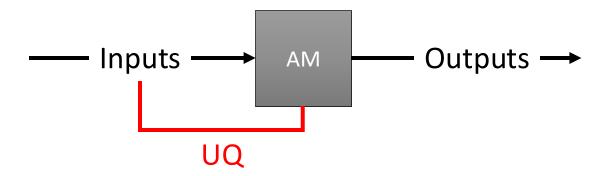
#### Melt pool images

- ~4770 image/layer
- 250 layers



### Verification, Validation, & Uncertainty Quantification





#### Command file

- Laser location
- Laser power
- Time

#### Data acquisition file

- Laser location
- Laser power
- Time

#### Melt pool images

- ~4770 image/layer
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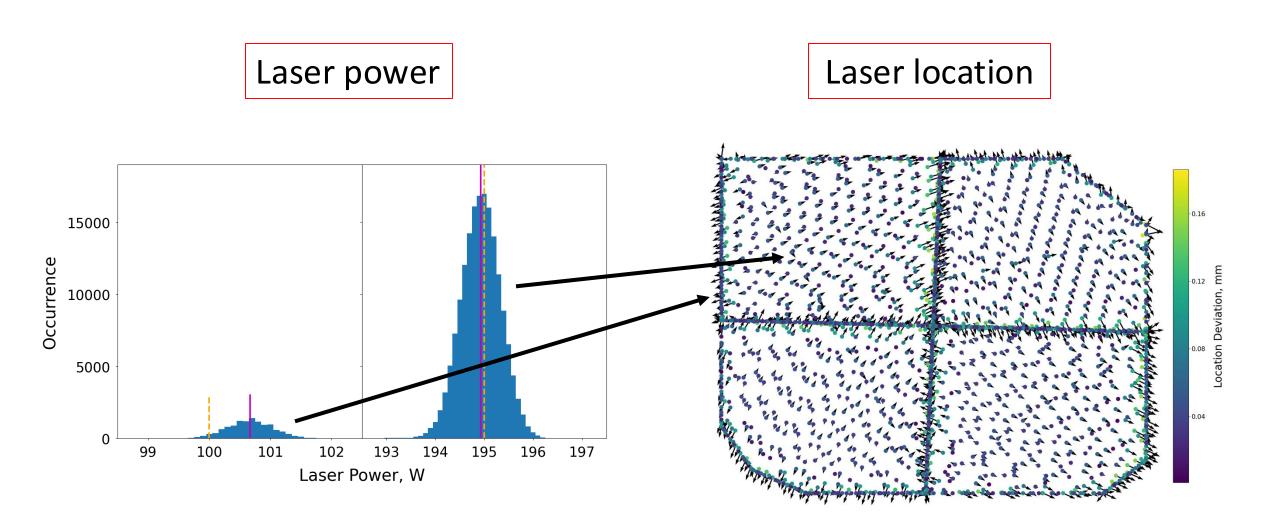
#### VVUQ to evaluate data quality

**UQ:** Uncertainty Quantification

SA: Sensitivity Analysis

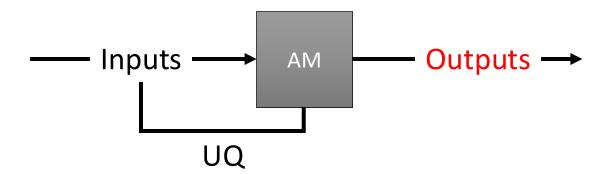
# Variability of the control





### Verification, Validation, & Uncertainty Quantification





#### Command file

- Laser location
- Laser power
- Time

#### Data acquisition file

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- Time

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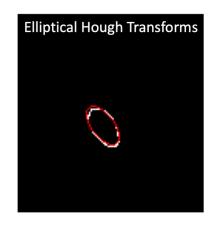
UQ: Uncertainty Quantification

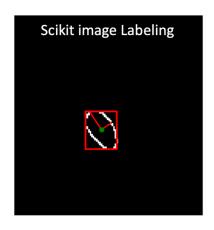
SA: Sensitivity Analysis

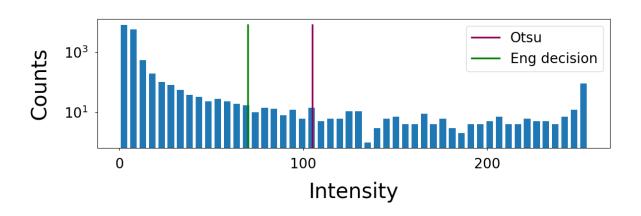
# Statistical Analysis & Feature Selection



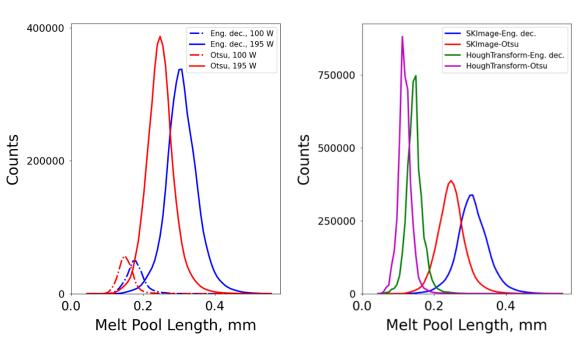






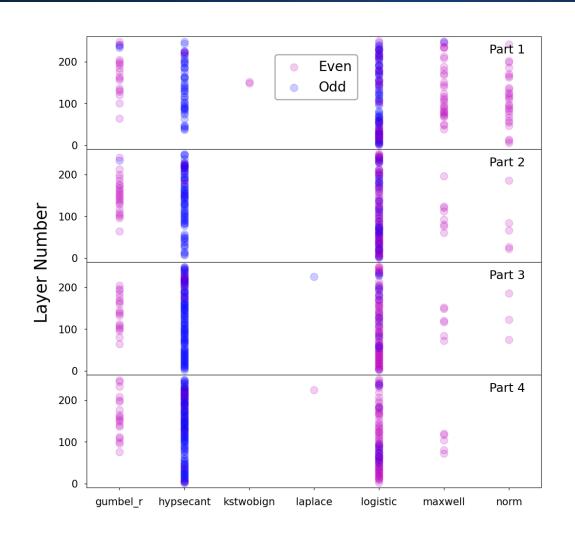


# Make images usable & define problem space

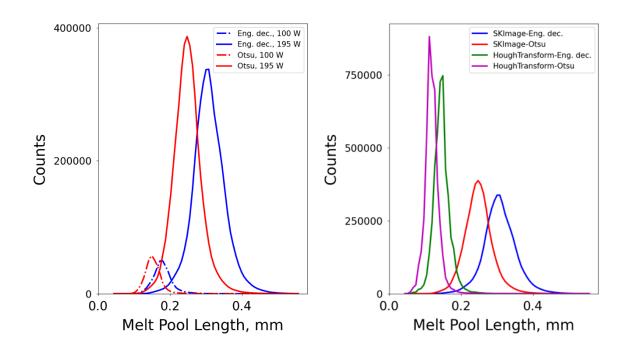


### Statistical Analysis & Feature Selection





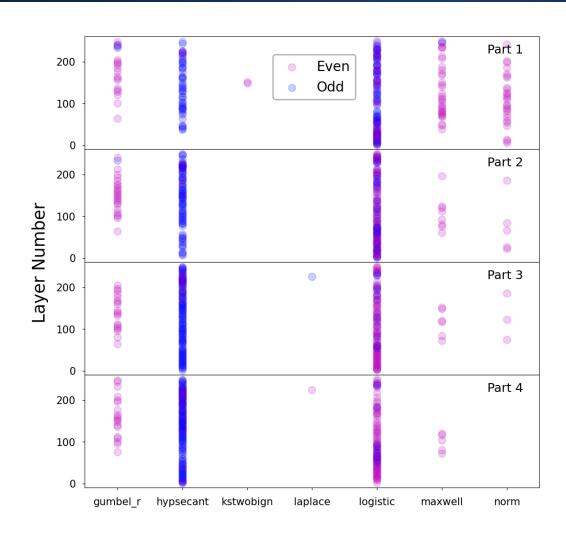
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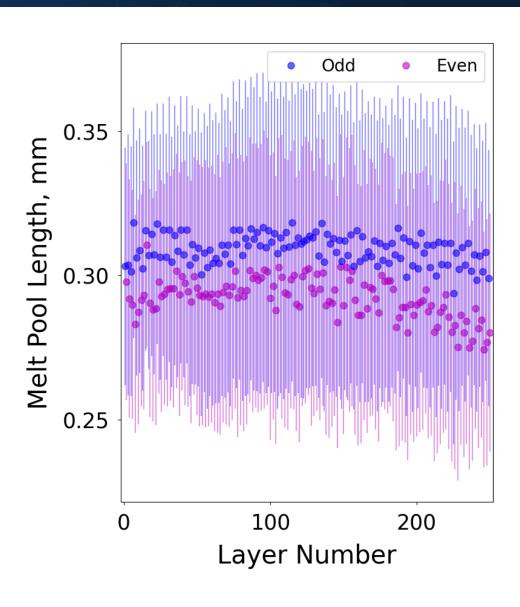
**Probability Distribution Function** 

# Statistical Analysis & Feature Selection



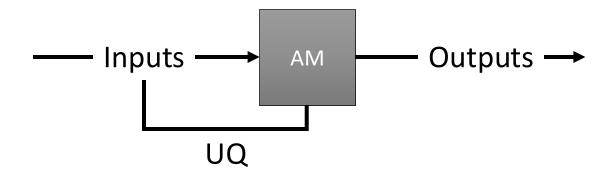


**Probability Distribution Function** 



### Verification, Validation, & Uncertainty Quantification





#### Command file

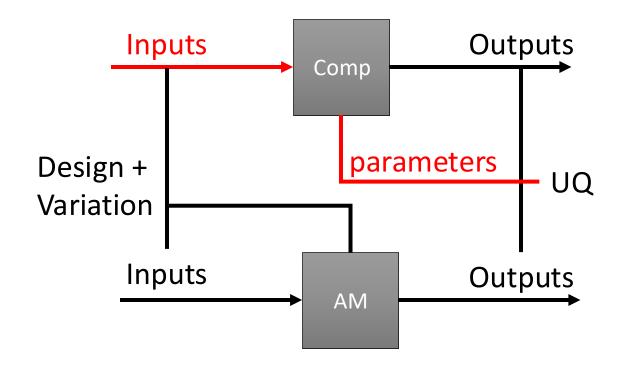
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- Laser location
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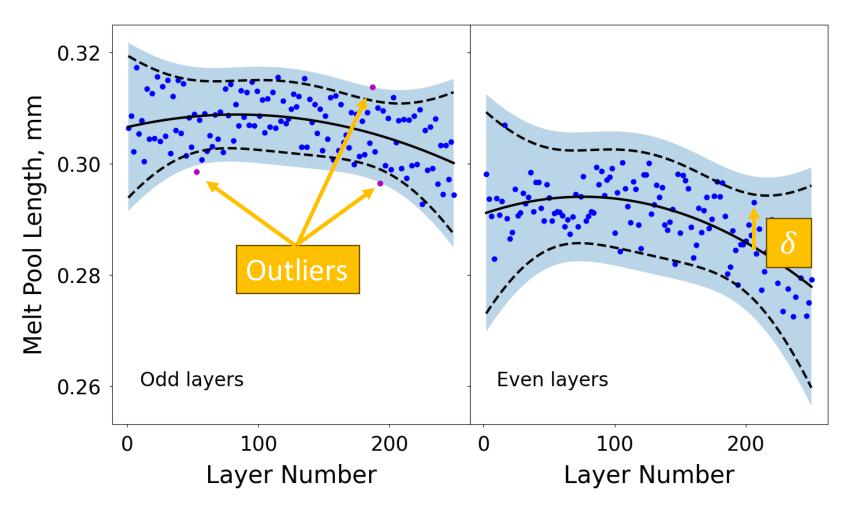


**VVUQ** to evaluate data quality

**VVUQ** to assess model quality

## Melt Pool Length Prediction

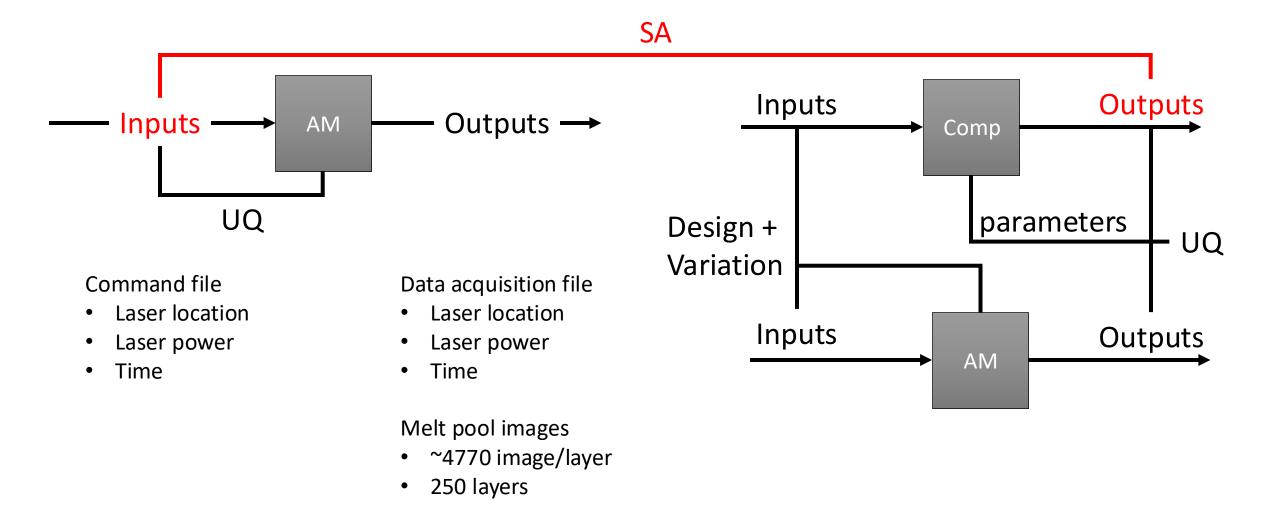




 $MPL = f_{2P}(layer\ height) + \delta$ Bayesian regression, PyMC3

### Verification, Validation, & Uncertainty Quantification



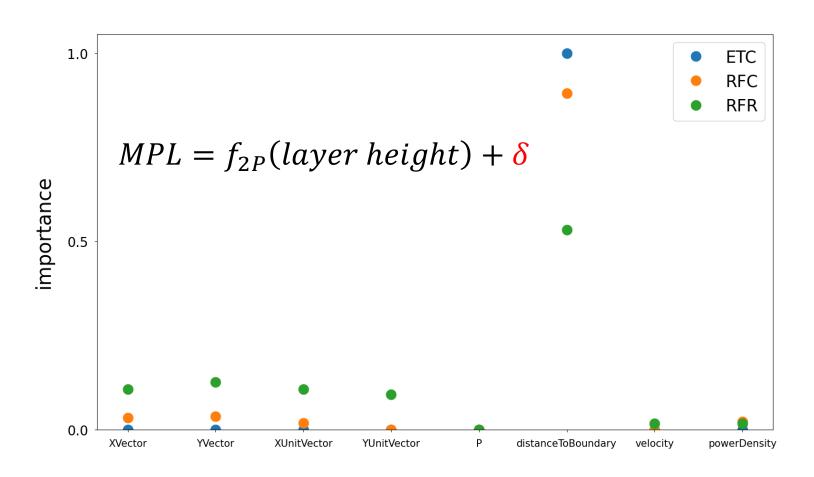


**VVUQ** to evaluate data quality

**VVUQ** to assess model quality

### Feature Selection





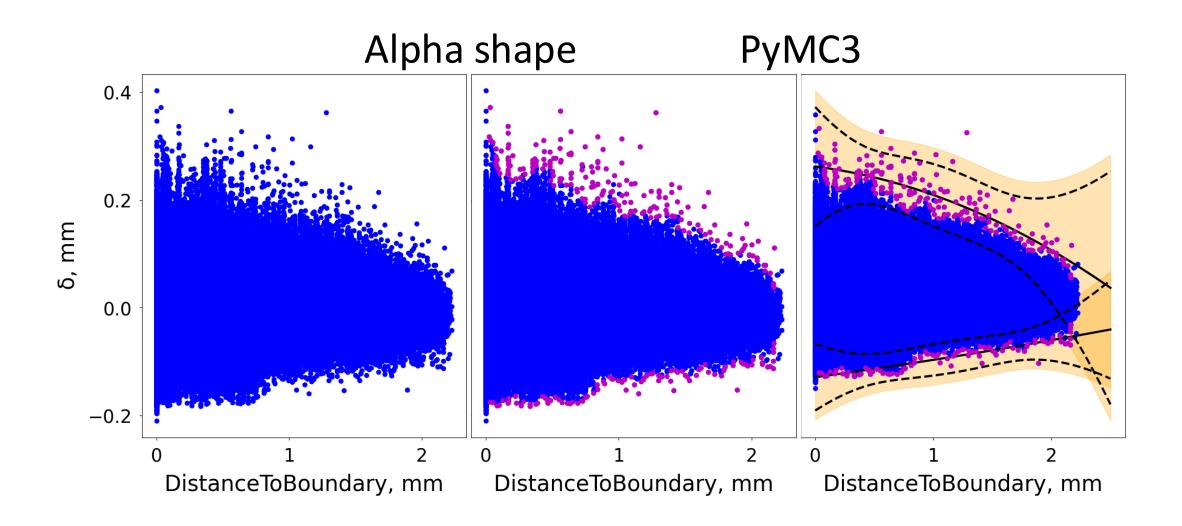
ETC: Extra-trees classifier

RFC: Random forest classifier

RFR: Random forest regression

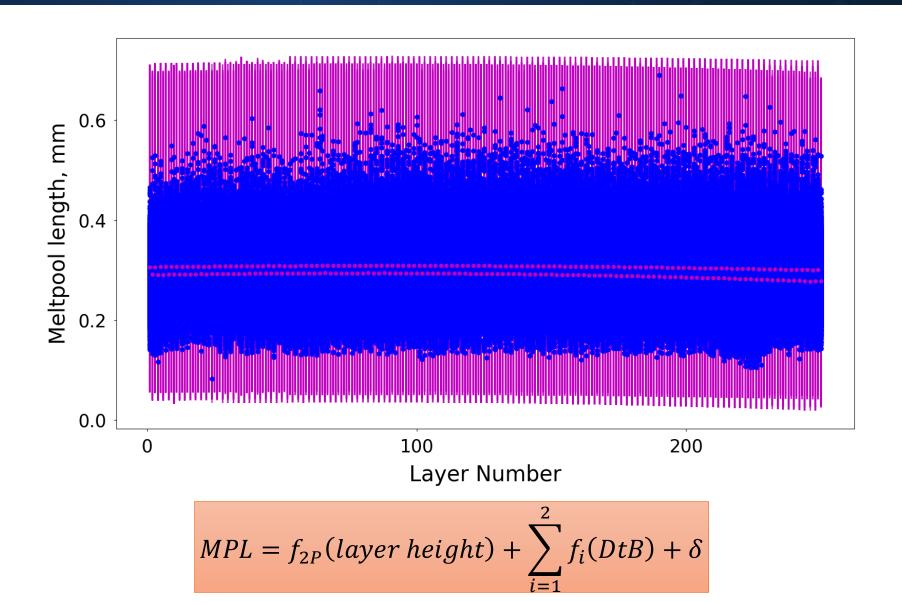
# **Uncertainty Quantification**





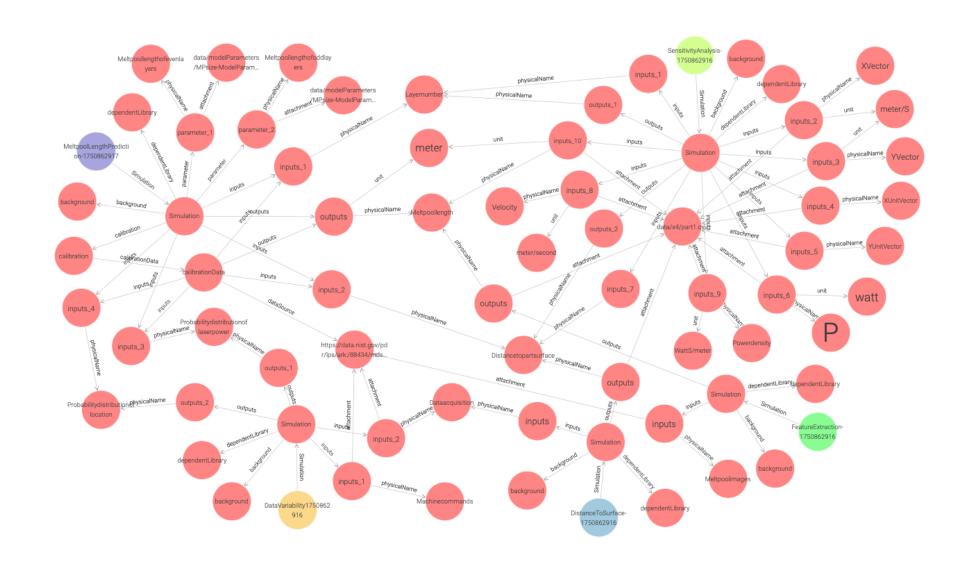
# Melt Pool Length Prediction





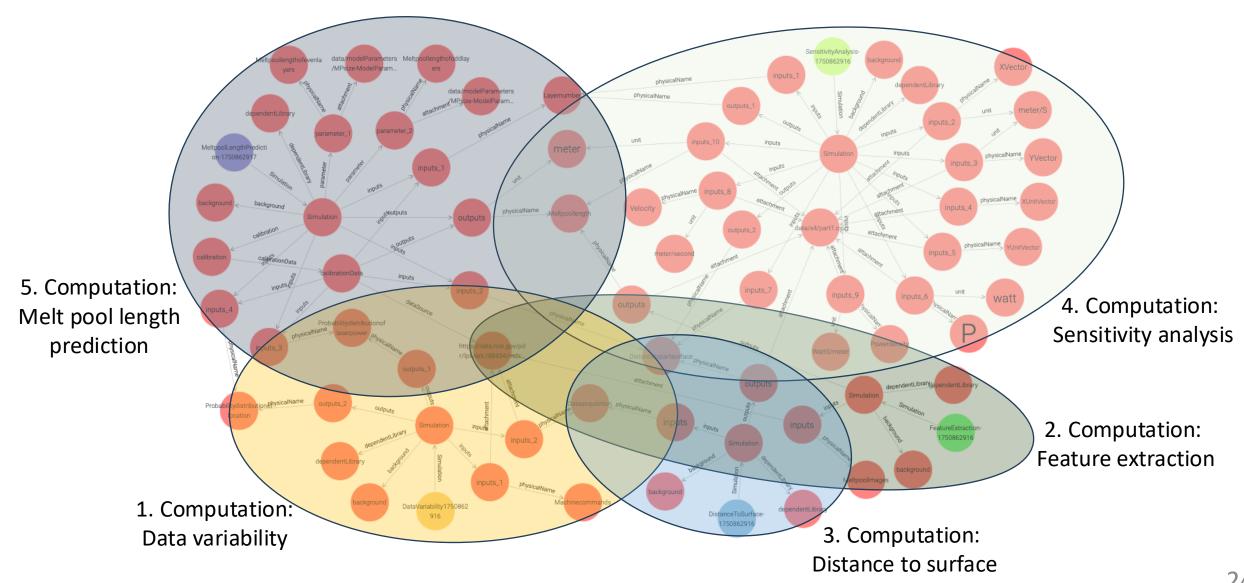
# Modeling Workflow





### Modeling Workflow – Reusable & Scalable





# Summary: Trustworthiness of Workflow



- A checklist is proposed to assess the trustworthiness of computational models at different phases of the lifecycle
- It enables the Reusability, Scalability, Transparency, and (Interoperability) of simulations

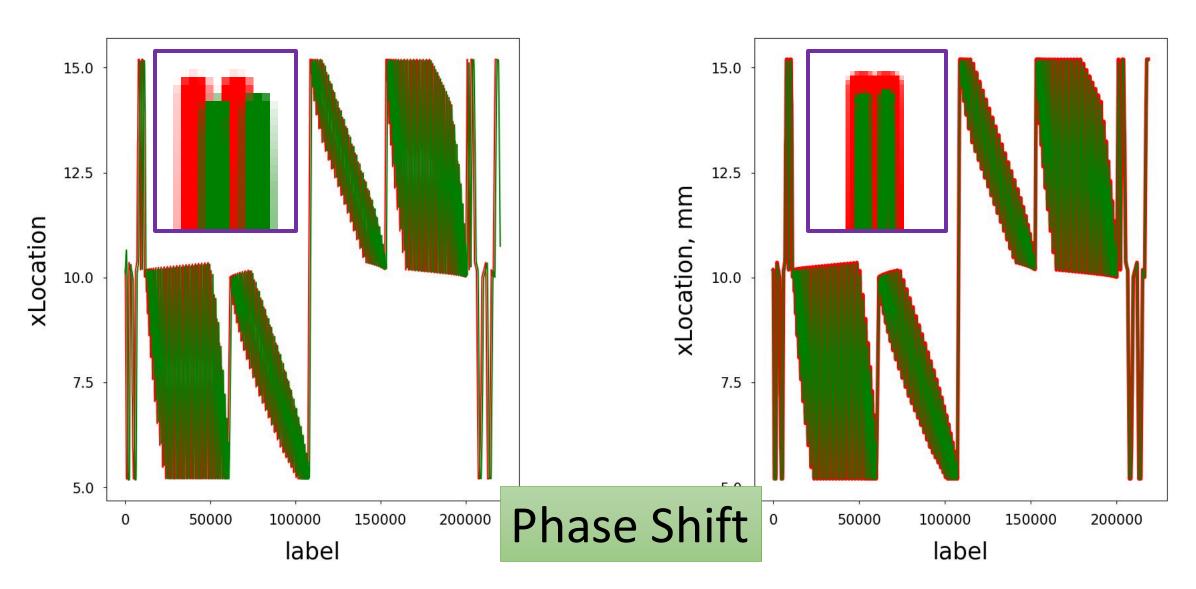
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Constraints	Uncertainty Quantification	Workflow Refineme	COLD PRODUCT COLD
	Computational Detail	Scalability	Fox.3 Fox.5 wat wat

shengyen.li@nist.gov



## **Data Preparation**





### Trustworthiness of Simulations

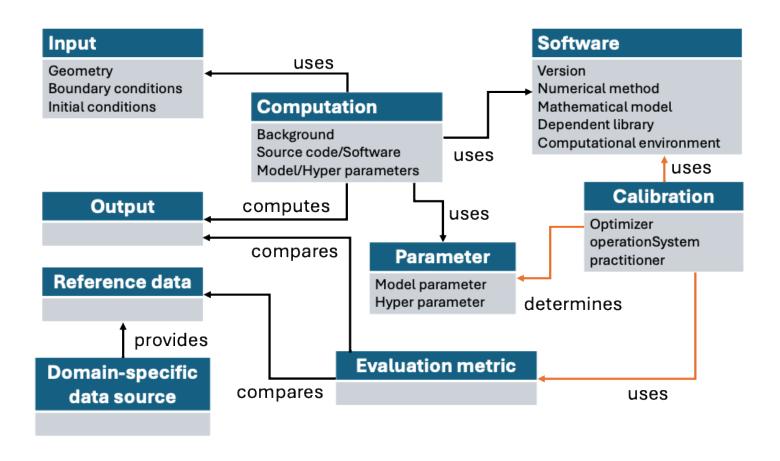


- Trustworthiness of simulations, hardware, integration and beyond
- A checklist to assess the trustworthiness of simulations in the 4 phases of the development lifecycle
- More specific information from ASME, ANS, IEEE, ISO, FDA, ...

Definition	Development	Deployment	Maintenance
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### Data Model for Computation

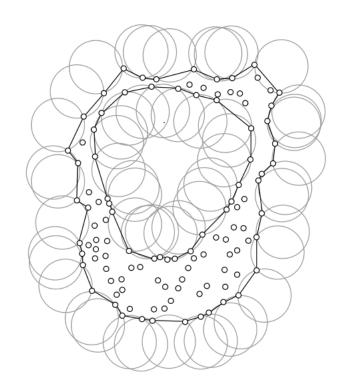




- A conceptual model has been implemented to address interoperability, reusability, and scalability.
- A Python library has been developed to translate the formats among XML schema, JSON, and JSON-LD.

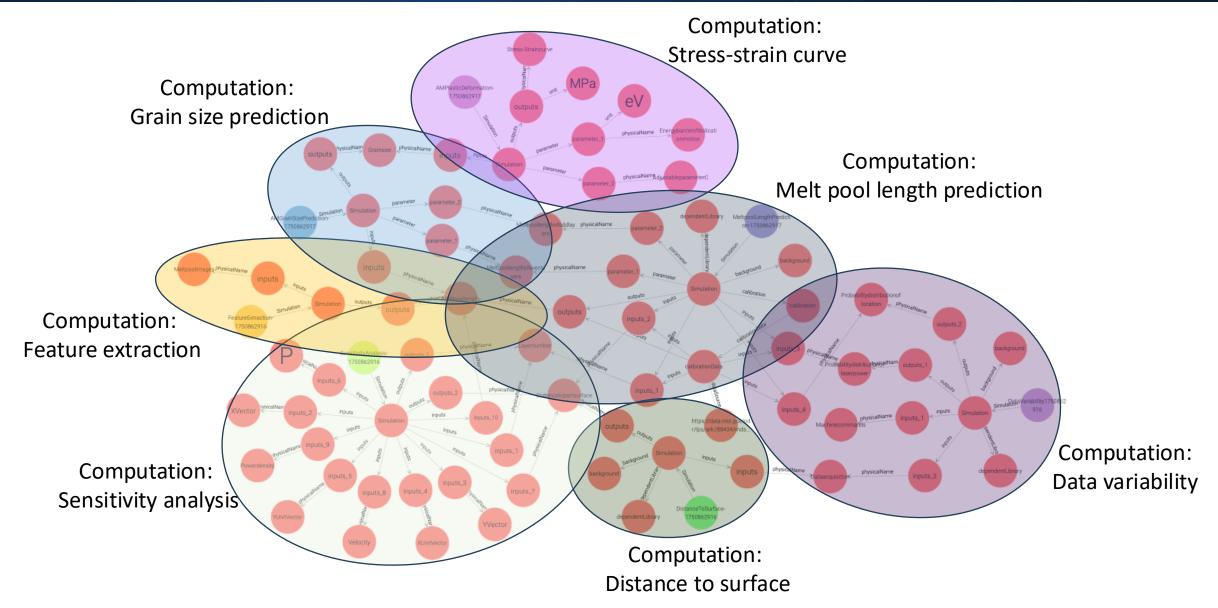


Alpha shape Introduction to Alpha Shapes, Kaspar Fischer



## Model integration example







#### 6 sigma:

Goal: 3.4 defects per million opportunities

### For Creation/Improvement

- 1. Define: Identify the problem and project goals.
- 2. Measure: Collect data to understand current performance.
- 3. Analyze: Identify root causes of defects.
- 4. Design/Improve: Implement and test solutions.
- 5. Verify/Control: Maintain improvements using monitoring and controls.

### Next steps



- Review and define vocabulary for computation
- Stress test this model using a material digital twin for additive manufacturing
- Iteratively validate this metadata model to improve the maturity level

