

### An Adaptive Geometry-Free Thermo-Mechanical Model for Directed Energy Deposition Process Modeling

September 2022

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# An Adaptive Geometry-Free Thermo-Mechanical Model for Directed Energy Deposition Process Modeling



### **Background & Outline**

#### □ Background

- Additive manufacturing & Directed Energy Deposition (DED) challenges
- Computational tools show promise for advancing the DED process
- Accurate & efficient computational model is required

#### □ Outline

- Material deposition model with mesh adaptivity
- Thermo-mechanical model
- Numerical examples & validation
- Recent advancements

#### Find the ideal materials

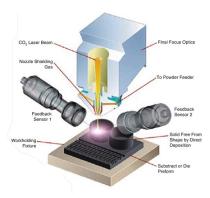
- discover metamaterial
- printability assessment



Challenges of DED (images are from *siemens.com* and *simplfy3d.com*)

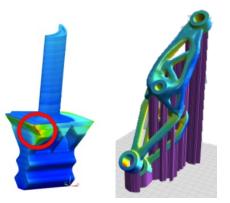
#### Qualify the process

- monitor and avoid print failures
- automate production



#### Create a perfect design

- optimize part designs
- optimize printing parameters



### **Material Deposition Model**

#### ■ Model material deposition based on subdomain construction

Decompose the physical domain:

$$\Omega = \Omega_s \cup \Omega_p, \quad \Omega_s \cap \Omega_p = \emptyset$$

Decompose the product subdomain:

$$\Omega_p = \Omega_a \cup \Omega_i, \quad \Omega_a \cap \Omega_i = \emptyset, \quad \Gamma_{a,i} = \partial \Omega_a \cap \partial \Omega_i$$

Move elements from inactive subdomain to the active subdomain:

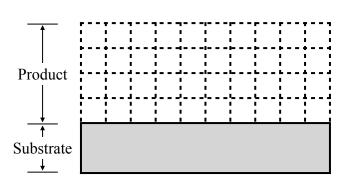
$$\Omega_i \leftarrow \Omega_i \setminus \{\mathcal{T}^i\}$$
 and  $\Omega_a \leftarrow \Omega_a \cup \{\mathcal{T}^i\}$ 

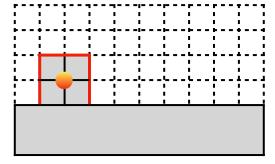
 $\square$   $\Omega_a$ 

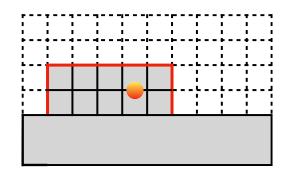
 $::::::\Omega_i$ 

 $---\Gamma_a$ 

Heat source







Initial step

First step

Second step

Element activation based on moving subdomains.

- Physical domain

 $\Omega_s$  - Substrate subdomain

 $\Omega_p$  - Product subdomain

 $\Gamma_{a,i}$ - Common interface

 $\Omega_a$  - Active product subdomain

 $\Omega_i$  - Inactive product subdomain

 $\mathcal{T}^{\iota}$  - Element in product subdomain

### **Material Deposition Model**

#### ☐ Element activation implementation

- Properties and initial conditions are applied to the newly activate elements
- Interface  $\Gamma_{a,i}$  is updated
- Boundary conditions (if any) are applied on the updated interface
- Based on block-restricted system in MOOSE



Multiphysics Object-Oriented Simulation Environment

#### □ Advantages

- Different definition of MOOSE-based objects in different subdomains
- Only part of the DoFs appear in the system of equations for a typical physics analysis
- Element activation & de-activation
- Open-source to all interested users

```
Algorithm 1: Element activation: (\Omega_a, \Omega_i, \Gamma_{a,i}) \leftarrow EA(\Omega_a, \Omega_i, \Gamma_{a,i}, \theta)

Initialization: activated element set \mathcal{A} = \emptyset;

for each element \mathcal{T}^{\iota} \in \Omega_i do

Compute average temperature in \mathcal{T}^{\iota}: \bar{\theta}^{\iota} = \left(\int_{\mathcal{T}^{\iota}} \theta \ d\Omega\right) / \left(\int_{\mathcal{T}^{\iota}} 1 \ d\Omega\right);

if \bar{\theta}^{\iota} > \theta_{melt} then

Add element to the active element set \mathcal{A} = \mathcal{A} \cup \{\mathcal{T}^{\iota}\};

If mesh adaptivity is utilized, move ancestors of \mathcal{T}^{\iota} to the active subdomain;

if \mathcal{A} \neq \emptyset then

Update subdomains: \Omega_a = \Omega_a \cup \mathcal{A}, \Omega_i = \Omega_i \setminus \mathcal{A};

Update interface and boundary information on \Gamma = \partial \Omega_a \cap \partial \Omega_i, \partial \Gamma_a, and \partial \Gamma_i;

Project boundary conditions, initial conditions, and material properties in all \mathcal{T}^{\iota} \in \mathcal{A};
```

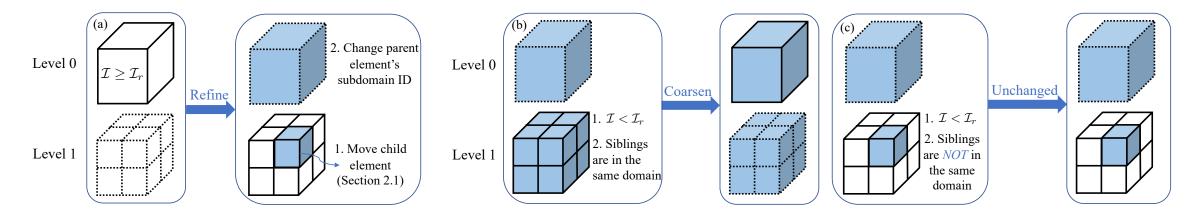
Key steps of the element activation process

### **Mesh Adaptivity**

#### ☐ Subdomain-consistent mesh adaptivity

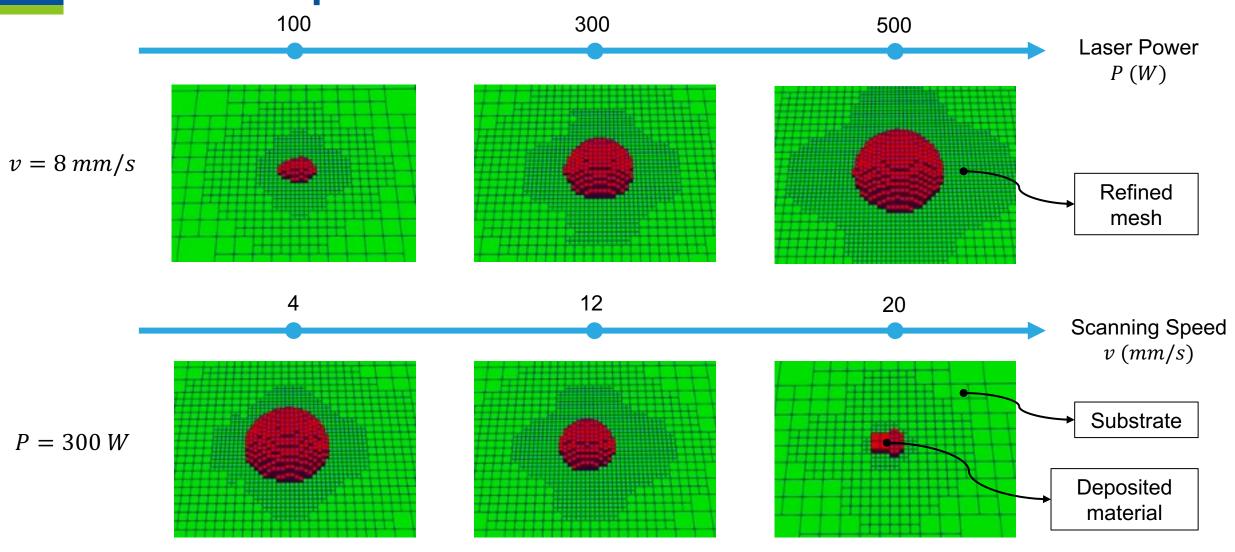
- Based on existing AMR in MOOSE and LibMesh
- Change parent element subdomain ID during refinement
- Only coarsen an element predicted error is above threshld & all siblings are in the same subdomain





- Schematic of the subdomain-consistent mesh adaptivity design. Only two refinement level is shown as an illustration.
- Highly resolved material interface representation
- Computationally efficient

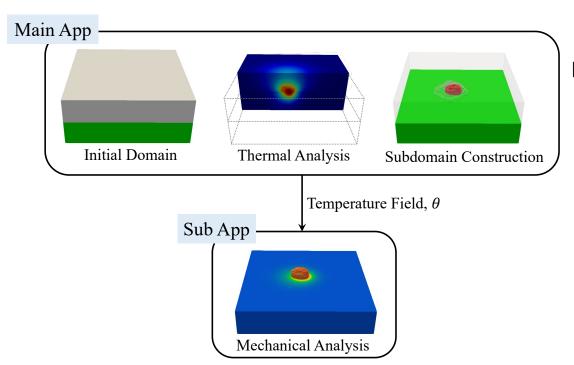
# **Material Deposition Demonstration**



### Thermo-mechanical Model – Finite Element Workflow

#### ■ Main steps

- 1. Solve the thermal governing equations in the entire domain
- 2. Mesh adaptivity is conducted based on the predicted temperature field error
- 3. Move elements that have averaged  $\theta > \theta_{melt}$  to the active product subdomain
- 4. Mechanical deformation due to thermal expansion is analyzed in the substrate and active product subdomain



#### ☐ Implementation details

- Multi-app capability
  - Main-app: thermal analysis, mesh adaptivity, material deposition
  - Sub-app: mechanical analysis
- Transfer capability
  - Communicate temperature field from main-app to sub-app
  - Mesh is kept identical between two apps

### Thermo-mechanical Model – Thermal Model

#### ☐ Thermal model

 Conservation of energy with a moving heat source and a convective boundary

$$\rho c(\theta) \frac{\partial \theta}{\partial t} = \nabla \cdot \kappa(\theta) \nabla \theta + Q(\boldsymbol{x}, t) \quad \text{in } \Omega,$$

$$\theta = \bar{\theta} \quad \text{on } \partial \Omega_{s, \text{bot}},$$

$$-\kappa(\theta) \nabla \theta \cdot \boldsymbol{n} = h(\theta) \cdot (\theta - \bar{\theta}_{\infty}) \quad \text{on } \Gamma_{a, i}.$$

#### ■ Moving heat source

Point heat source:

$$\hat{Q}(\boldsymbol{x},t) = \frac{2\alpha\eta P}{\pi r^3} \exp\left\{-\frac{2||\boldsymbol{x} - \boldsymbol{p}(t)||^2}{r^2}\right\}$$

• Line heat source:

$$\bar{Q}(\boldsymbol{x},t) = \frac{1}{\Delta t} \int_{t_0}^{t_0 + \Delta t} \hat{Q}(\boldsymbol{x},t) dt$$

Hybrid heat source:

$$Q(\boldsymbol{x},t) = egin{cases} \hat{Q}(\boldsymbol{x},t) & ext{if } \Delta l < L \ ar{Q}(\boldsymbol{x},t) & ext{if } \Delta l \geq L \end{cases}$$

 $\theta$  - Temperature

 $\rho$  - Density

 $c(\theta)$  - Specific heat ( $\theta$  dependent)

 $\kappa(\theta)$  - Thermal conductivity ( $\theta$  dependent)

 $h(\theta)$  - Heat convection coefficient ( $\theta$  dependent)

 $ar{ heta}$  - Room temperature

 $ar{ heta}_{\infty}$  - Temperature far from the boundary

P - Laser power

 $\alpha\,$  - Scaling factor

 $\eta$  - Laser efficiency

r - Effective radius of laser beam

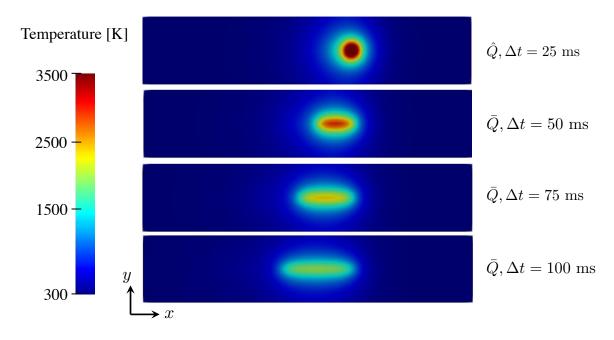
 $oldsymbol{p}(t)$  - Center of laser beam (scanning path)

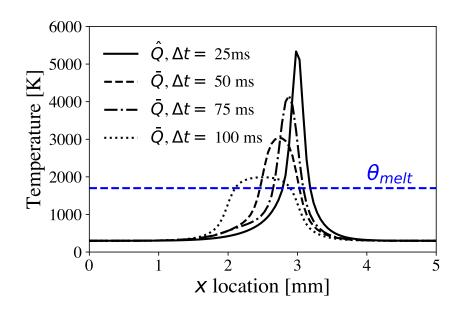
 $L\,$  - Threshold length

 $\Delta l$  - Laser beam moving distance during one time step

### **Thermo-mechanical Model – Heat Source Demonstration**

#### ☐ Heat source comparison





- Temperature fields of the Gaussian point and line heat sources with different time step sizes.
- Temperature profile as a function of the x location for the Gaussian point and line heat sources with different time step sizes.
- **❖** Hybrid heat source allows for a larger time step size than the point heat source
- **❖** Hybrid heat source is more accurate than the line heat source

### Thermo-mechanical Model – Mechanical Model

#### ☐ Mechanical model

Quasi-static conservation of momentum

$$\nabla \cdot \boldsymbol{\sigma} + \boldsymbol{b} = 0 \quad \text{in } \Omega_{a,s}$$
$$\boldsymbol{u} = \boldsymbol{0} \quad \text{on } \partial \Omega_{s,\text{bot}}$$

Power-law hardening elastic-plastic constitutive relation

$$oldsymbol{\sigma} = egin{cases} E( heta)oldsymbol{arepsilon}, oldsymbol{\sigma} \leq oldsymbol{\sigma}_y \ Koldsymbol{arepsilon}^n, oldsymbol{\sigma} > oldsymbol{\sigma}_y \end{cases}, \quad oldsymbol{\sigma}_y = \left(rac{E^n( heta)}{K}
ight)^{1/(n-1)}$$

Additively decompose the total strain

$$\varepsilon = \varepsilon_e + \varepsilon_p + \varepsilon_\theta$$

Strain due to thermal expansion

$$\boldsymbol{\varepsilon}_{\theta} = \beta(\theta - \theta_0)\boldsymbol{I}$$

$$\theta_0 = \begin{cases} \bar{\theta}_{\infty} & \text{in } \Omega_s \\ \theta_{\text{melt}} & \text{in } \Omega_a \end{cases}$$

 $\sigma$  - Rank-two Cauchy stress tensor

 $oldsymbol{b}$  - Body force

 $oldsymbol{u}$  - Displacement

 $E(\theta)$ - Young's modulus ( $\theta$  dependent)

K - Strength coefficient

n - Strain hardening coefficient

 $oldsymbol{\sigma}_y$  - Yield stress

 $\varepsilon$  - Total strain

 $oldsymbol{arepsilon}_e, oldsymbol{arepsilon}_p, oldsymbol{arepsilon}_{ heta}$  - Elastic, plastic, and thermal strain

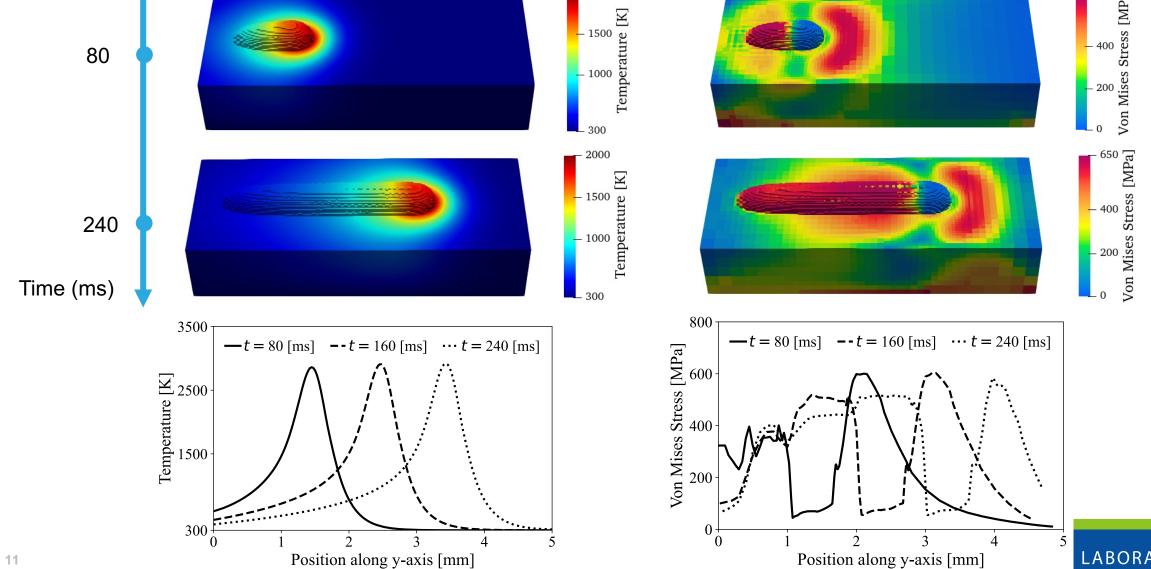
eta - Thermal expansion coefficient

 $\theta_0$  - Stress-free temperature

 $heta_{
m melt}$  - Melt temperature

I - Rank-two identity tensor

### **Single Track Example**



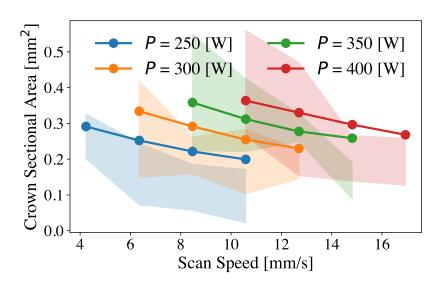
### Single Track Example - Validation

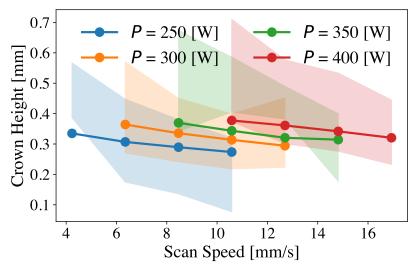
#### □ Problem setting

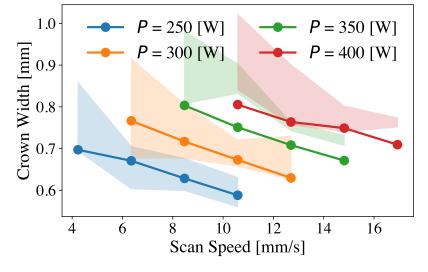
- Single track, single layer, scan from 0.5mm 3.5mm
- 4 laser powers, 4 scanning speeds
- Idaho National Laboratory's Optomec MTS500 LENS DEDtype laser/blown- powder 3D metal printer system

Laser Power (W)	Effective Radius (mm)	Scanning Speed (mm/s)
250	0.27	4.23, 6.35, 8.47, 10.58
300	0.30	6.35, 8.47, 10.58, 12.70
350	0.33	8.47, 10.58, 12.70, 14.81
400	0.37	10.58, 12.70, 14.81, 16.93

#### Simulation parameters for each case







Comparison of simulated and experimental material bead feature sizes

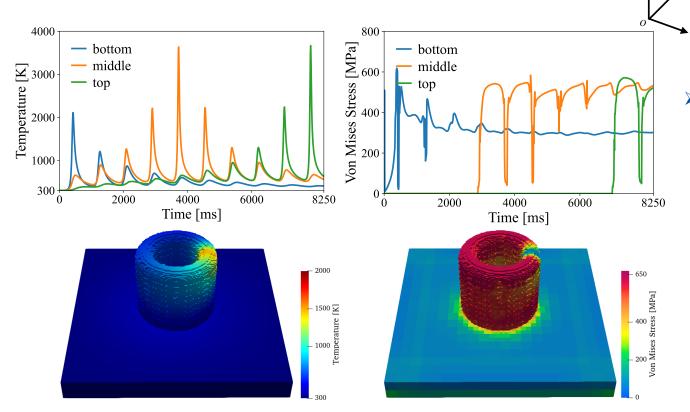
# **Hollow Cylinder Example**

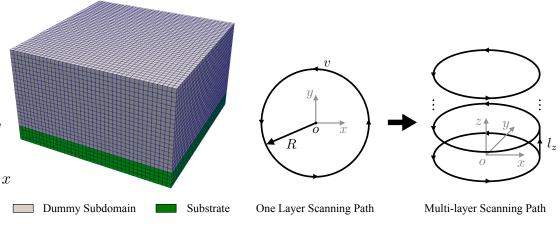
#### □ Problem setting

• Domain size  $10 \ mm \times 10 \ mm \times 6 \ mm$ , initial mesh size  $0.25 \ mm$ 

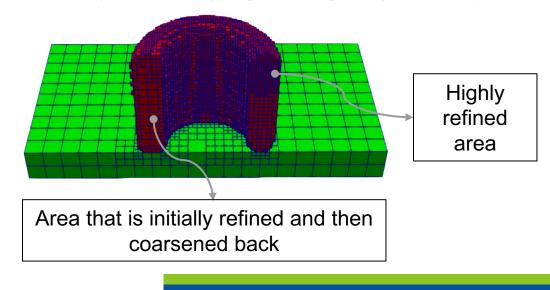
• Scanning path radius R = 1.5 mm, vertical spacing  $l_z = 0.3 mm$ 

• Laser power 250 W, scanning speed 11.43 mm/s





Problem setting for simulating the process of printing a hollow cylinder



# Recent Advancements – Heat Source Improvement

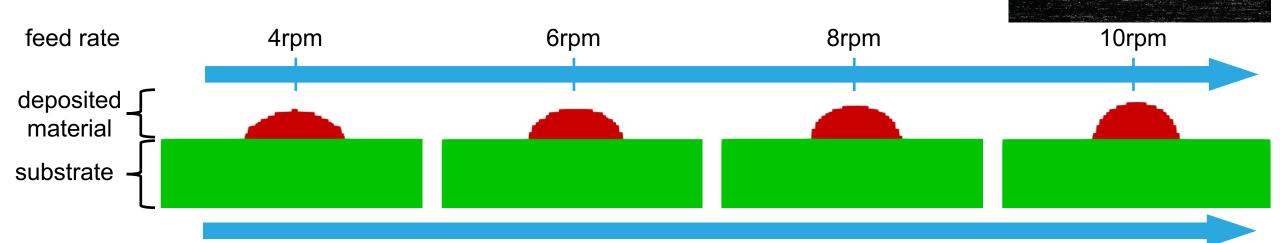
#### ☐ Effects from the change of feed-rate

Anisotropic point heat source:

$$\hat{Q}(x, y, z, t) = \frac{2\alpha\eta P}{\pi r_x r_y r_z} \exp\left\{-2\left(\frac{x - p_x(t)}{r_x^2}\right)^2 - 2\left(\frac{y - p_y(t)}{r_y^2}\right)^2 - 2\left(\frac{z - p_z(t)}{r_x^2}\right)^2\right\}$$

Effective radii follows normal distributions :

$$r_x, r_y \sim \mathcal{N}\left(\mu_1, \sigma_1^2\right); \quad r_z \sim \mathcal{N}\left(\mu_2, \sigma_2^2\right) \quad \mu_1, \mu_2, \sigma_1, \sigma_2 \text{ are fitted from experimental data}$$



# Recent Advancements – Automatic Height Tracking

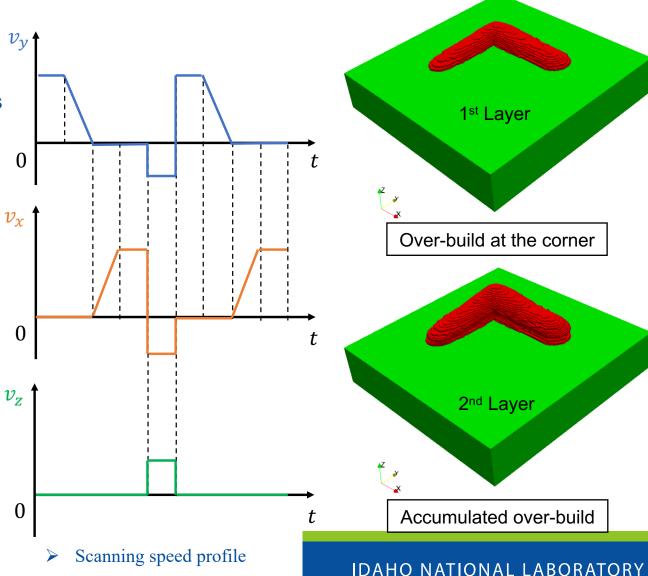
#### $\square$ In-situ adjustment of heat source height $p_z(t)$

- Capture overbuild during manufacturing
- Capture overbuild accumulation for higher buildups
- Heat source height  $p_z(t)$  is not an input parameter, but the maximum z-coordinate of the built material in the  $(p_x(t), p_y(t))$  plane





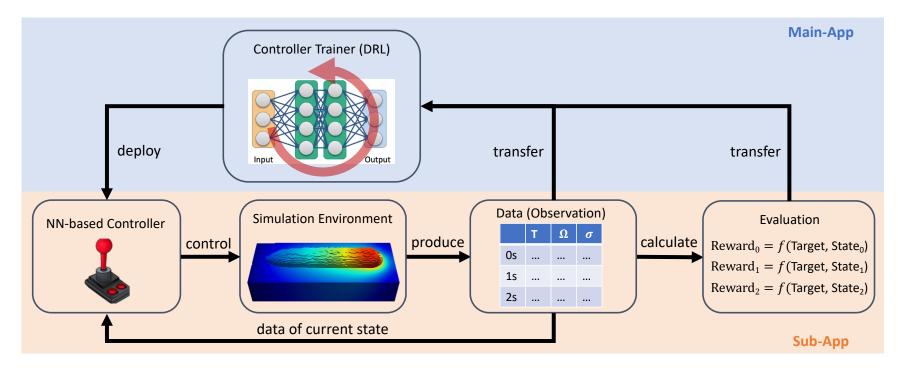
Sub-optimal built parts (over-build along the circumference and at the corners)



### Recent Advancements – Control and Optimization

#### ☐ Al-based process control and optimization

- Develop the neural-network based machine learning capabilities in MOOSE (integration of MOOSE & PyTorch)
- Develop the deep reinforcement learning (DRL)-based control scheme
- Research into DRL algorithms, control workflow



Schematic of the AI-based process control and optimization implementation

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