



Image-guided Additive Manufacturing

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Industrial relevance

The proposed project will address defects, improve build quality, reduce costs, and increase the yield of additive manufacturing.

Facts:

- 😊 **Customization** – easier to build complex, highly customized parts, add more flexibility in the design and manufacturing.
- 😊 **Rapid growth of market** – \$4 billion in 2015, \$10.8 billion in 2021.
- 😞 **High rejection rate (>2%)** – various types of defects induced by process variations (e.g., thermal effects and extraneous noises).
- 😞 **Long post-build inspection (~25% of the manufacturing time)** – low yield (e.g., for engine parts) and high cost.

Urgent need: In-situ process monitoring and control.

Thrust area: Intelligence

Current TRL: TRL-4

Final TRL: TRL-8

Project type: Proposed project

Percent complete: 30%

Problem statement

Motivation

Process variations (e.g. laser or galvo instabilities or drift, thermal effects, variations in powder feedstock) perturb **build quality** and generate **residual stress** → the distortion or embedded flaws → impact **mechanical properties** such as fatigue strength.

Gaps

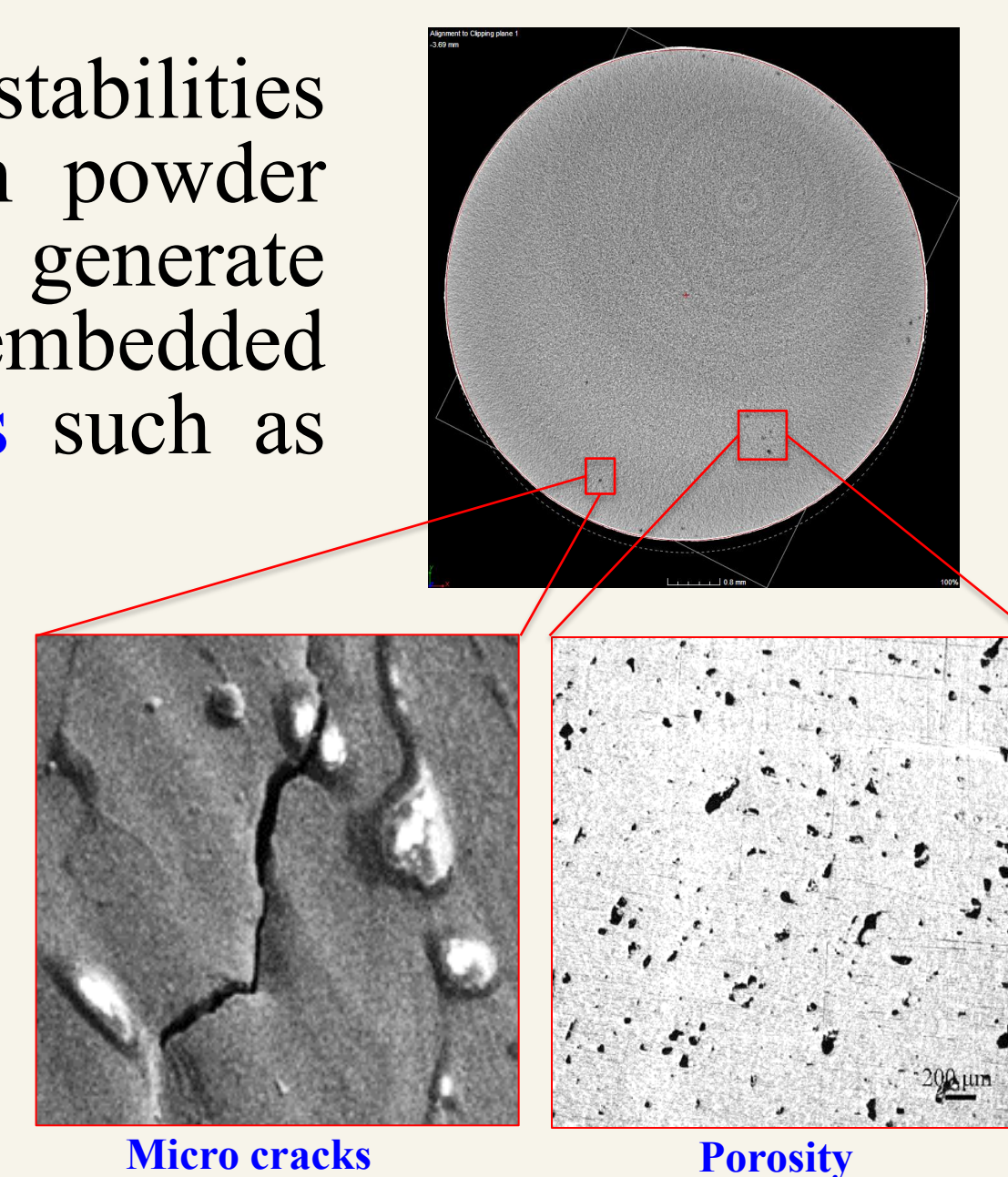
- Depend on **post-build inspection** or **destructive tests**.
- Lack the ability to extract pertinent information about process dynamics from images.

Need

Quality control and **process repeatability** are critical to mass use of additive manufacturing.

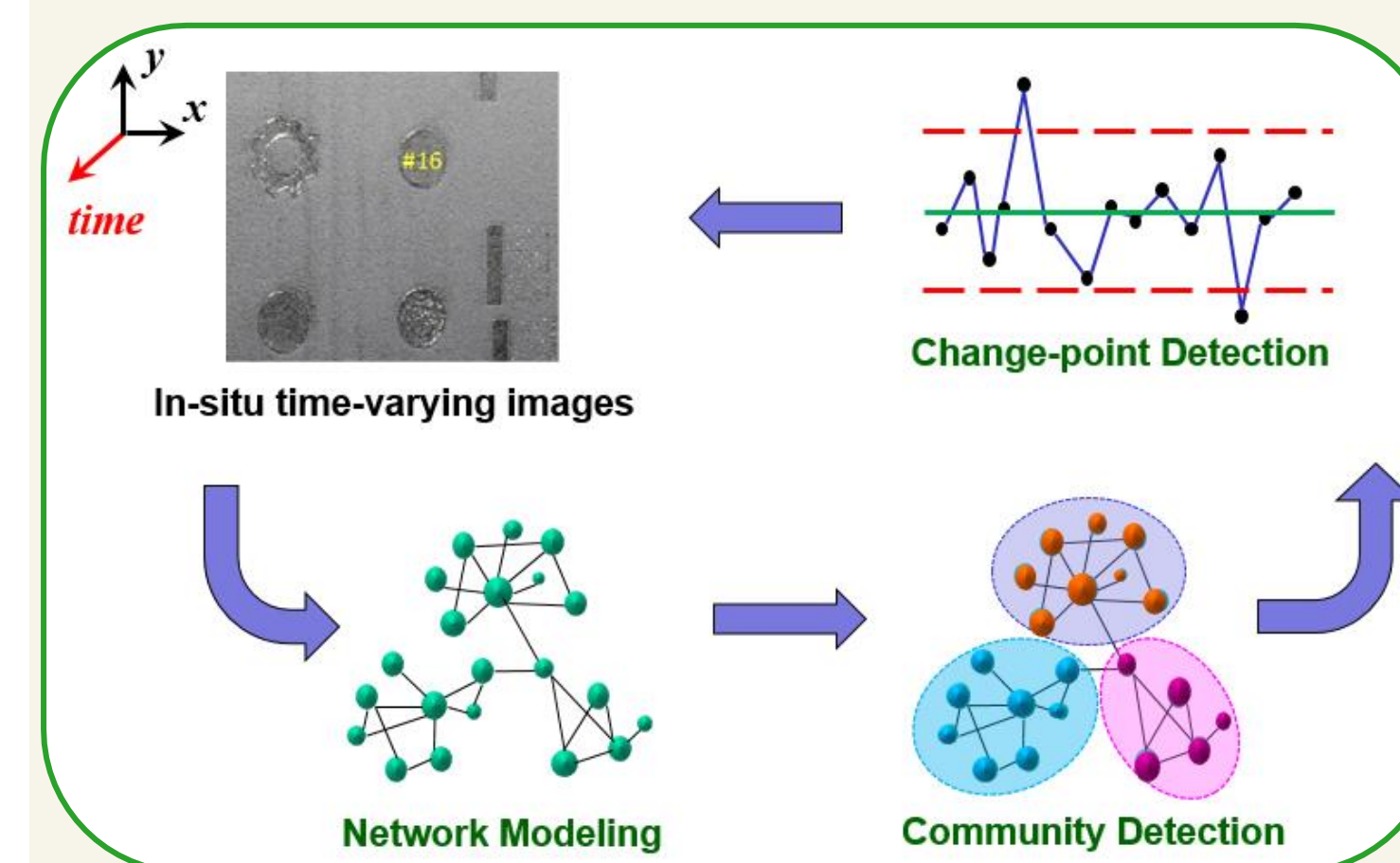
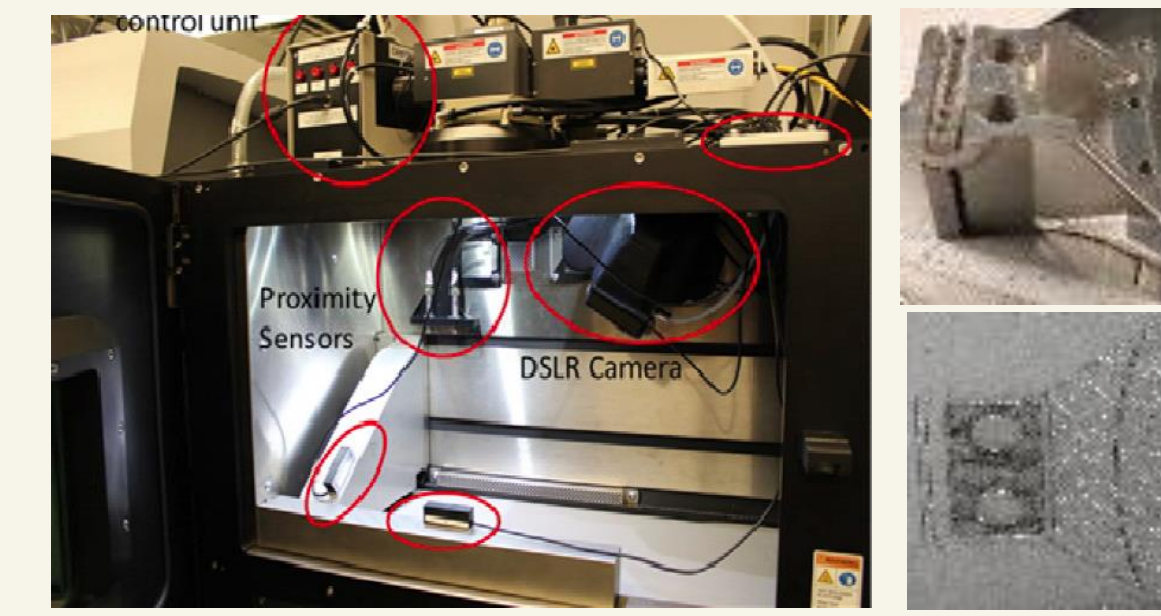
Objective

Develop an **in-situ image sensing, fusion and decision-support system** for **real-time defect mitigation** in additive manufacturing.



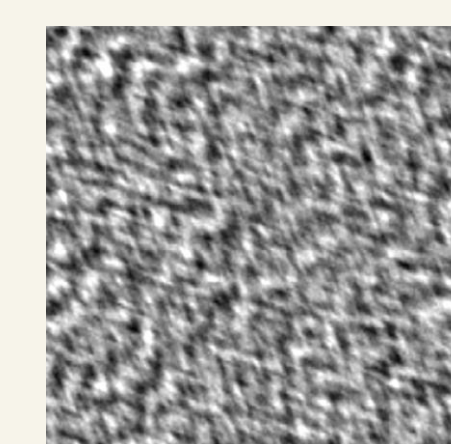
Approach and method

- In-situ image sensing (high-resolution **DSLR camera**, **infrared thermography** and **CT scanner** with >1000 Hz frame rate)

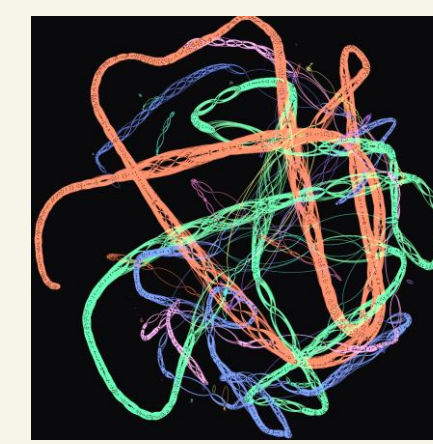


- **Network GLR chart** for real-time detection of defects

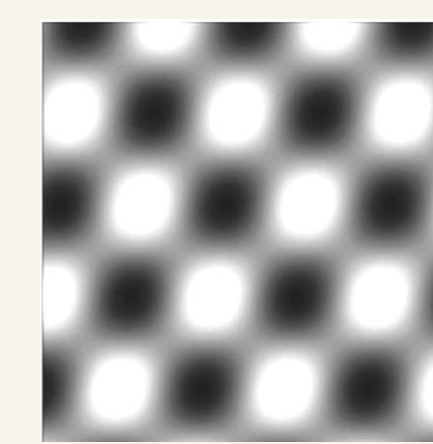
- Characterization of **network variations** → address defects



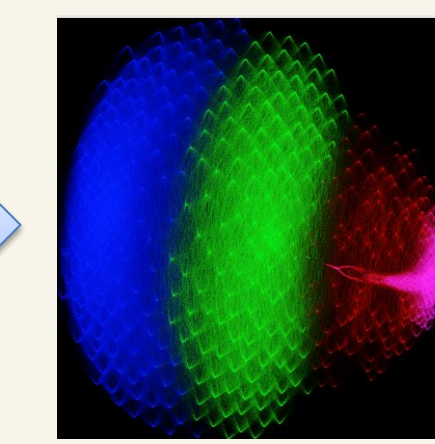
Auto regression



Network



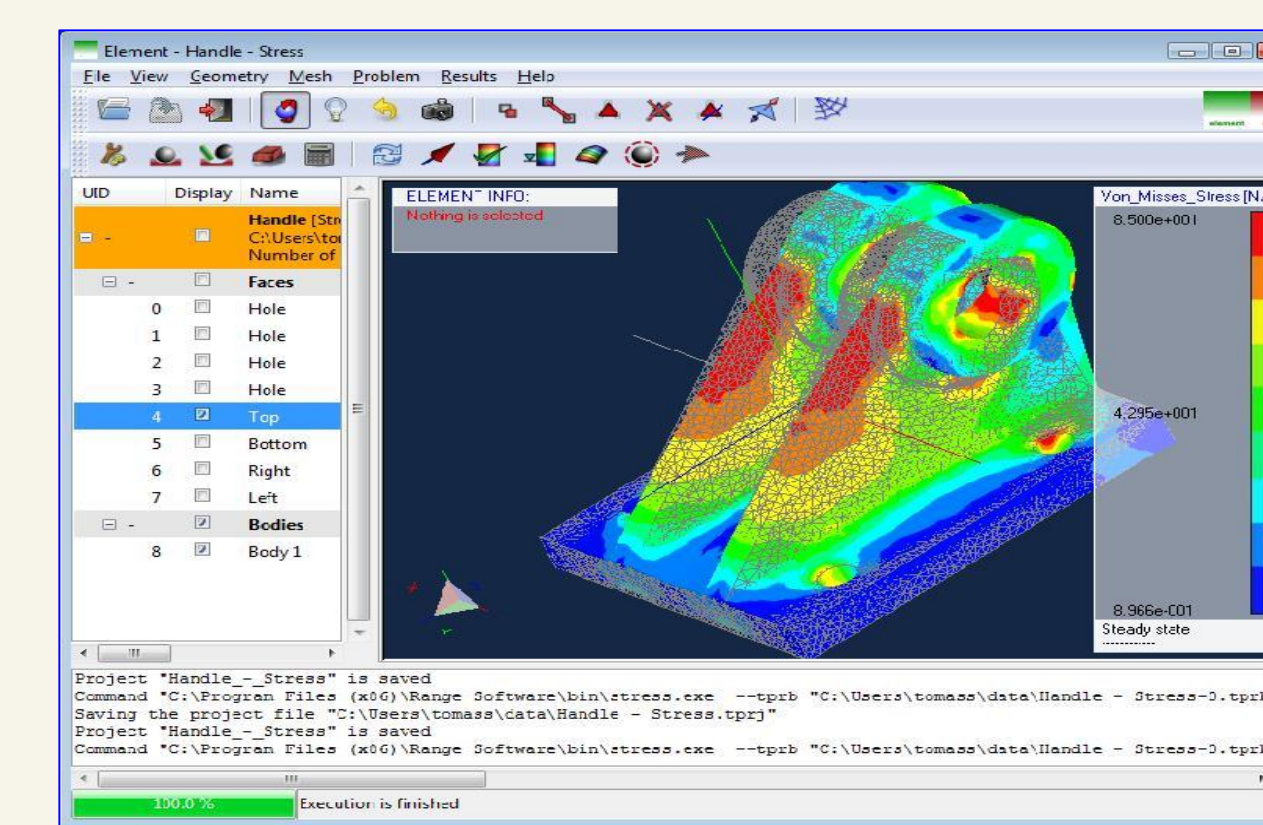
Periodic



Network

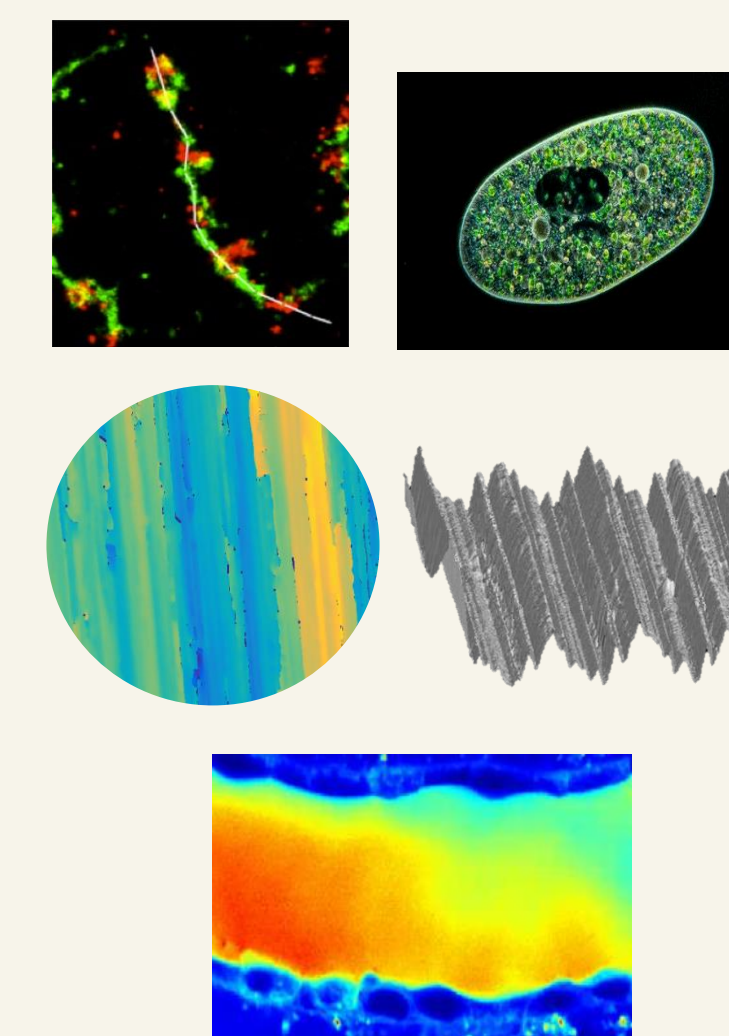
Deliverables and benefits

- An **in-situ image sensing system** that will record in-process images and CT scans of every single layer of the product while it is built.
- A **software package with graphical user interface** that will extract pertinent features about process dynamics from images for the real-time defect mitigation and optimal control.



Potential application areas

- **Medical research** – anomaly detection in medical images (e.g., tumor detection).
- **Semi-conductor manufacturing** – defect inspection in microscopic images for assessment of surface finish of wafers.
- **Bio-manufacturing** – change detection in cell images for in-vivo monitoring of cellular processes.



Project plan and progress

- Review current literature, data cleaning and pre-processing.
- Extract features from image data to characterize process.
- Develop data-driven models to derive quantitative relationships between image features and defect evolutions.
- Develop image-guided control policy of additive manufacturing.
- Evaluation and validation

	Month 1-2	Month 3-4	Month 5-6	Month 7-8	Month 9-10	Month 11-12	End
Literature review							
Data cleaning							
System development							
o Image feature extraction							
o Data-driven models							
o Image-guided control							
Evaluation and validation							
Document and presentation							

Current state of practice and research

- Spectral band selection approach (Du *et al.* 2007) – analyze hyperspectral images of poultry carcasses.
- Low-rank tensor decomposition + multivariate control chart (Yan *et al.* 2015) – monitoring sequential flame images.
- Adaptive Gaussian process (Zhang *et al.* 2015) – characterize wafer thickness.
- Multistage semi-classification approach (Park *et al.* 2013) – characterize morphology of nanoparticles.
- Spatiotemporal control chart (Megahed *et al.* 2012) – detect non-conforming tiles.

Gap: *Very little work has been done to develop new image-guided additive manufacturing.*

How ours is different

Our proposed approach

- Just-in-time vs. Inventory
- In-process control vs. Post-build inspection
- High-dimensional images vs. Low-dimensional quality variables
- Dynamic image streams vs. Static images

Existing practices

References

- Chen Kan and Hui Yang, “Dynamic network modeling of in-situ image profiles for process control – applications in ultraprecision machining and biomanufacturing processes”, *submitted*, 2016.
- Chen Kan and Hui Yang, “Network models for monitoring high-dimensional image profiles”, *proceeding of IEEE International Conference on Automation Science and Engineering (CASE)*, pp. 2078-2083, Aug. 24-28, 2015, Gothenburg, Sweden.