

In-Situ Thermal Imaging for Quality Classification of Dissimilar Metal Laser Welds in Battery Tab Welding

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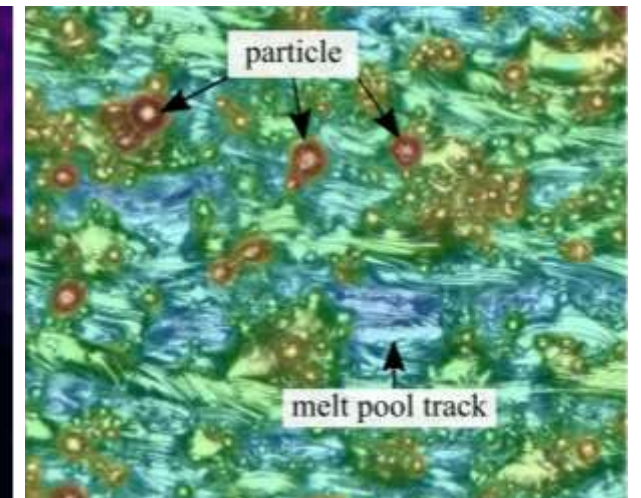
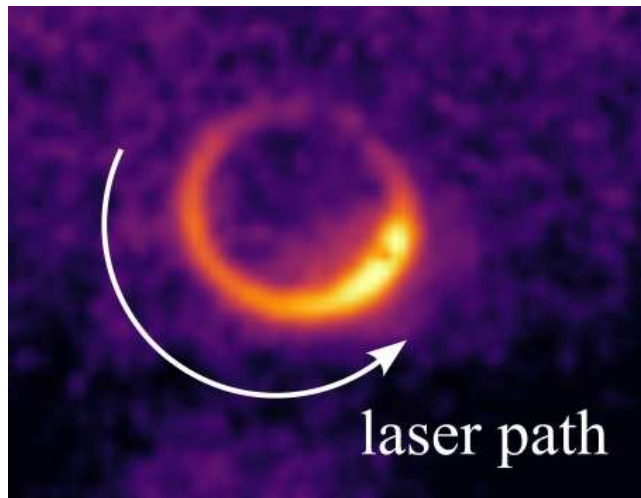
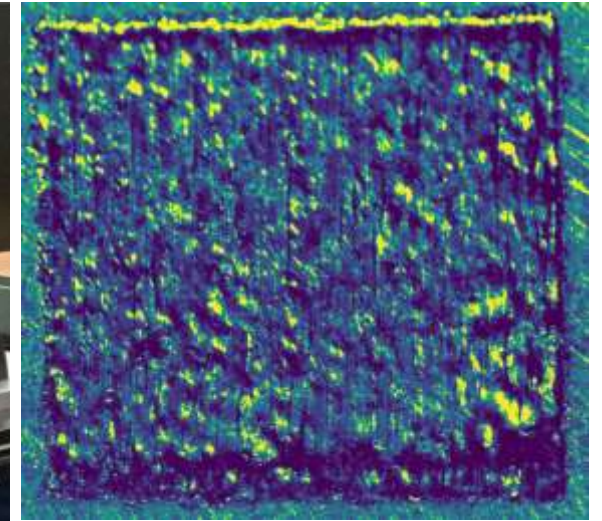
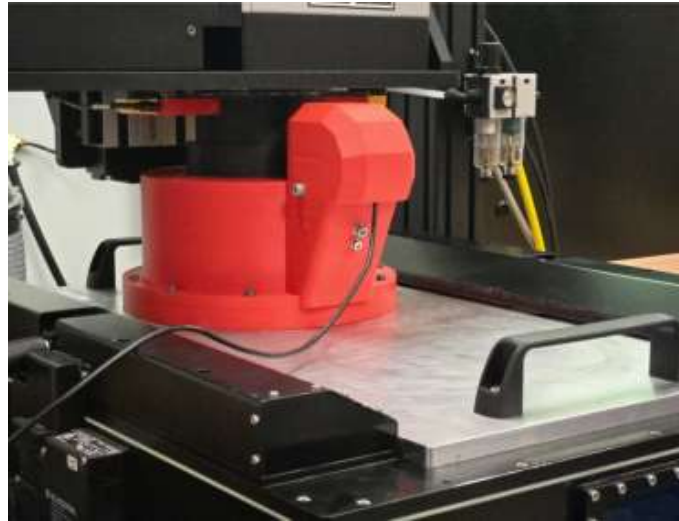
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Laser Powder Bed Fusion (LPBF)

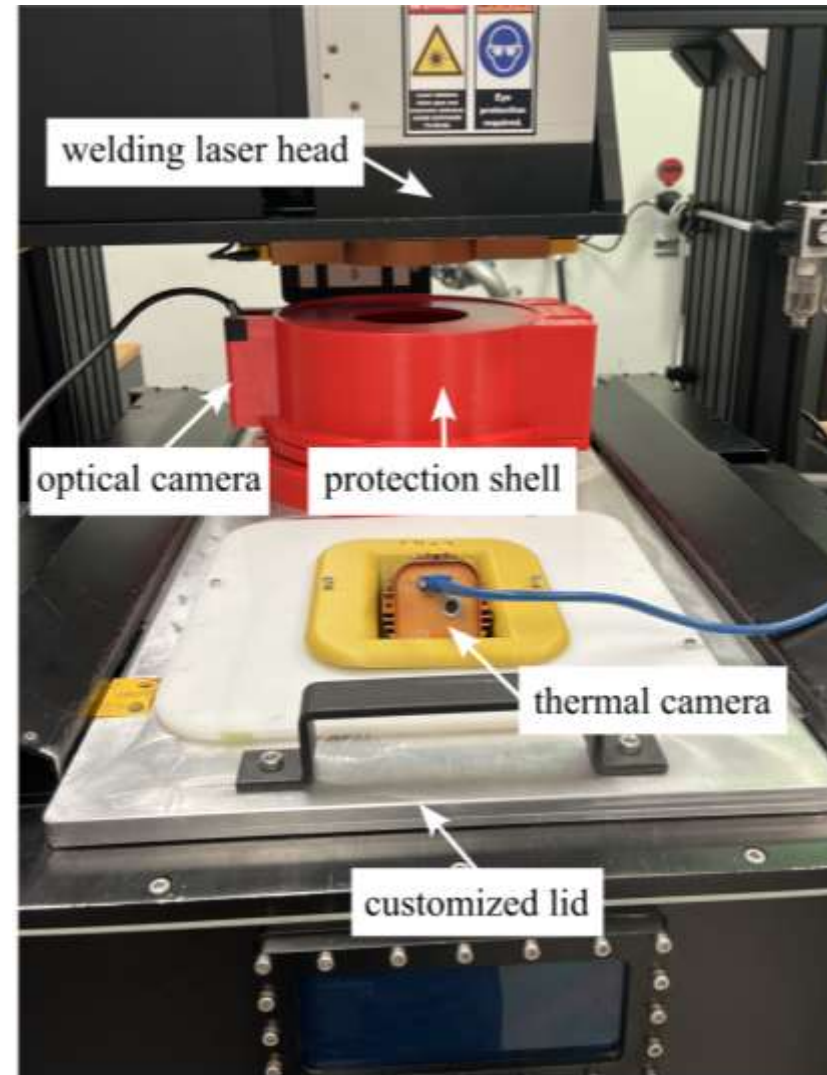
- Surface Quality and Melt Pool Behavior
- Prior Work on In-Situ Process Monitoring
- Camera Module Design for Inter-layer LPBF Monitoring



Prior Work on In-Situ Process Monitoring

Experimental Setup for LPBF Process

- LPBF system instrumented with optical and thermal imaging
- Controlled build using 316L stainless steel samples
- Part: 10 mm × 10 mm
- Material: 316L stainless steel
- Printing parameters:
 - Laser power: 200 W
 - Scan speed: 100 mm/s
 - Laser spot size: 100 μm
 - Hatch distance: 100 μm



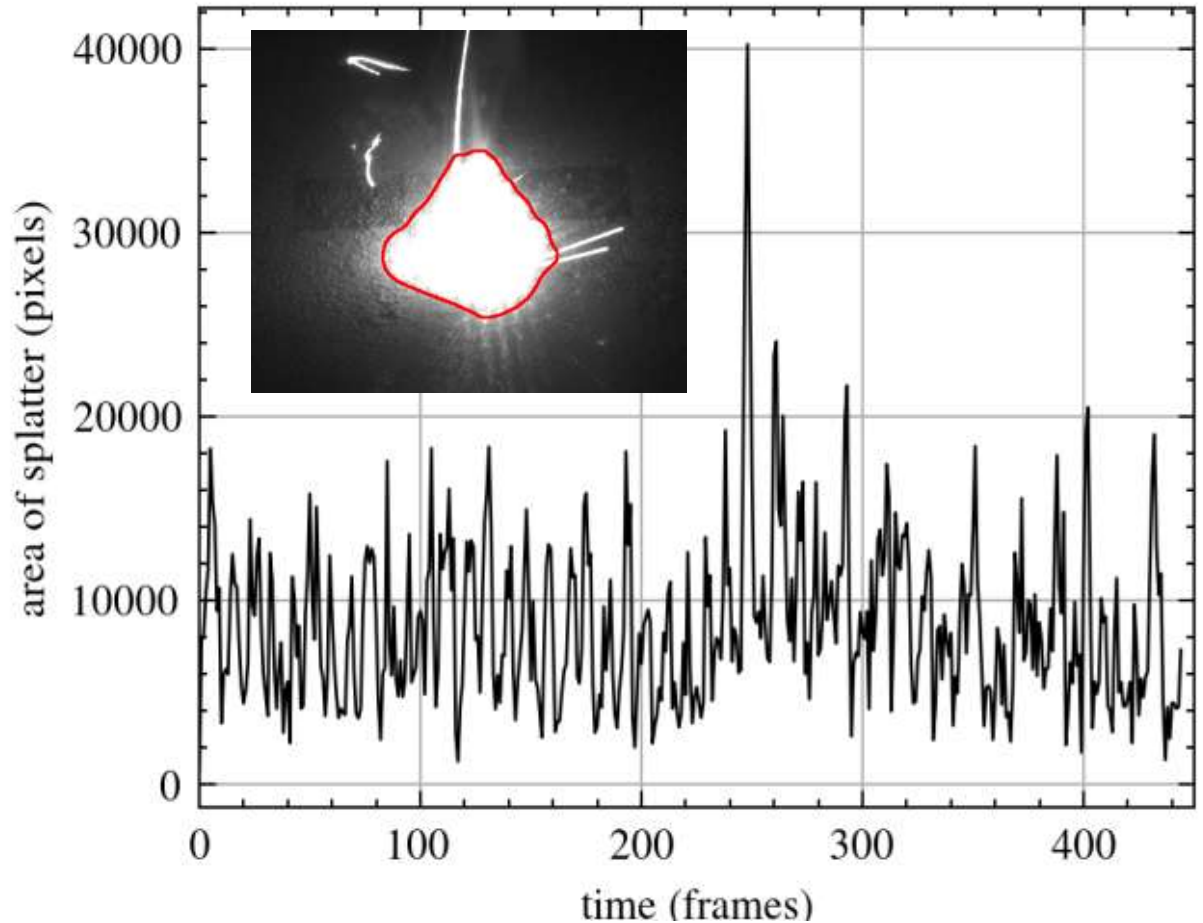
Imaging of LPBF Welding Process

- Captured using a FLIR Blackfly USB optical camera (1440 × 1080 resolution, 100 fps).
- Tracks melt pool, plume, and spatter dynamics in real-time.
- MidOpt BP660 bandpass filter (640-680 nm) enhances imaging clarity.



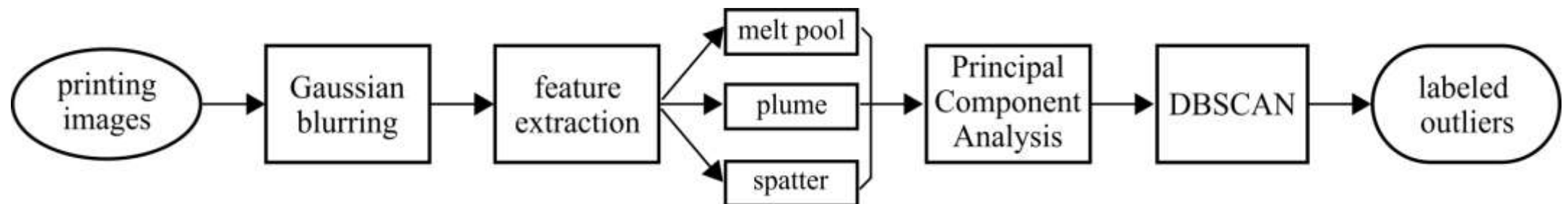
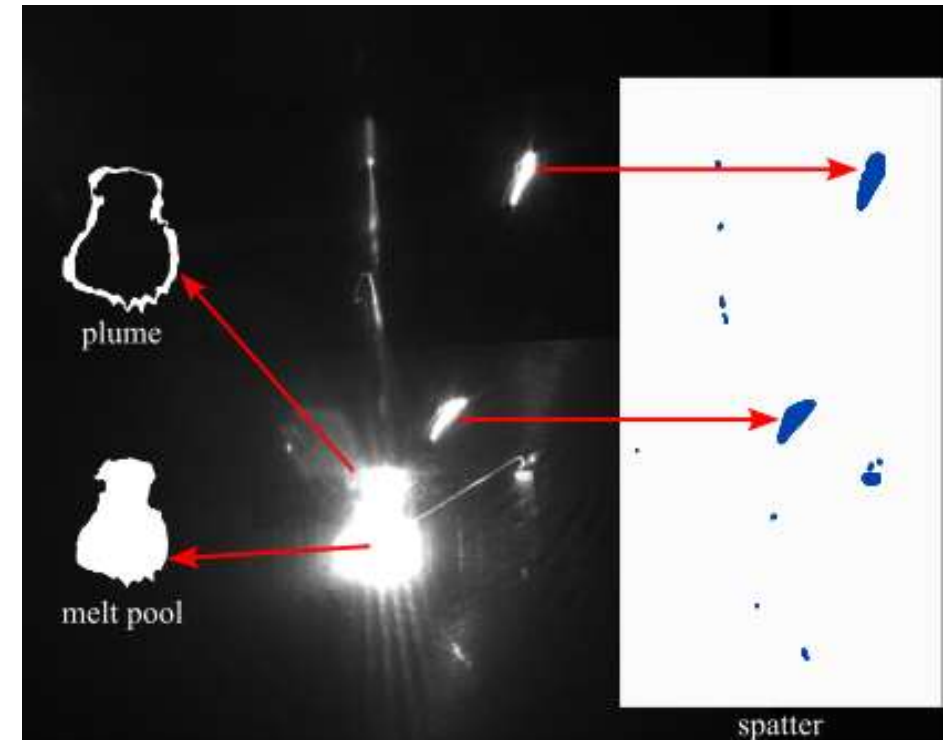
Real-Time Weld Feature Tracking

- Image processing used to track the weld/plasma region
- Tracks properties such as area, intensity, and motion
- Captures overall weld behavior, not limited to spatter alone
- Enables real-time monitoring with reduced data complexity



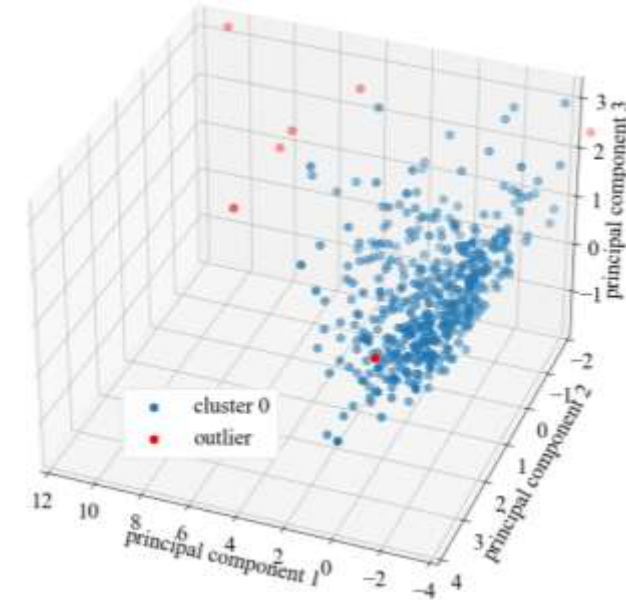
Feature-Based Process Monitoring

- Multiple features extracted: melt pool, plume, and spatter
- Image preprocessing improves feature segmentation
- Dimensionality reduction (PCA) captures key process variations
- Enables data-driven analysis of printing behavior



Unsupervised Defect Detection (DBSCAN)

- DBSCAN clustering separates normal and abnormal process states
- Outliers correspond to potential defect formation
- No labeled data required for training
- Demonstrates potential for real-time anomaly detection

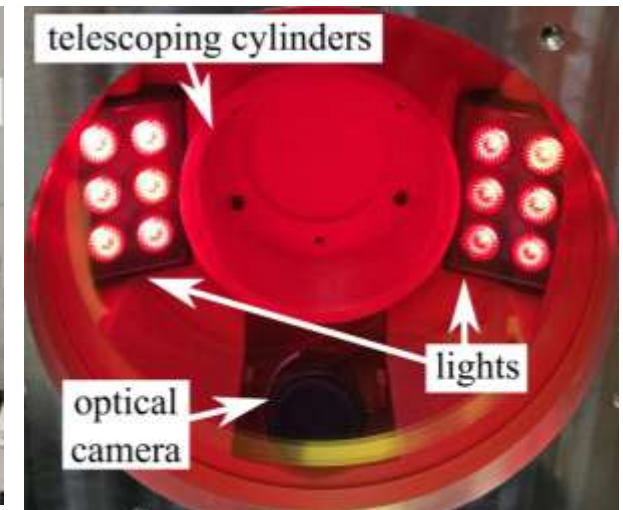
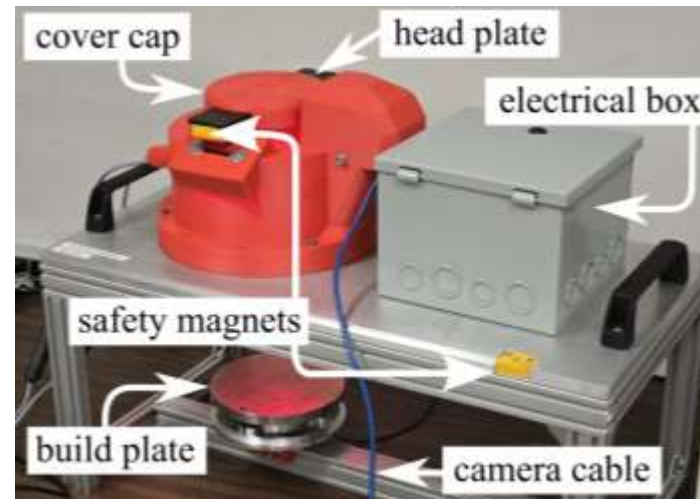
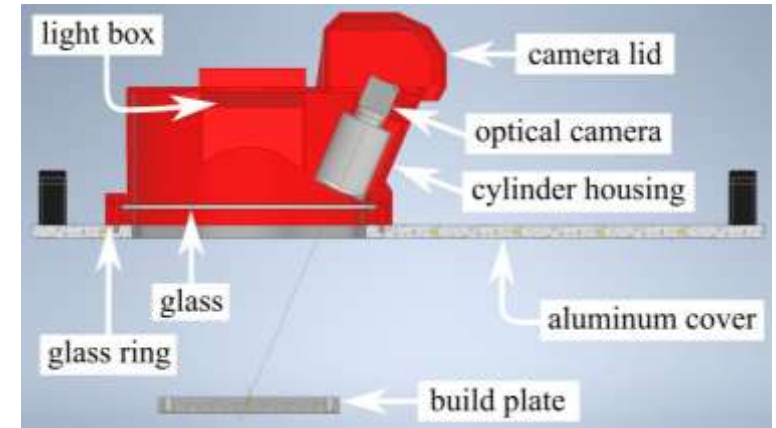


Yanzhou Fu, Matthew Whetham, Austin R. J. Downey, Lang Yuan, and Gurcan Comert. A study of online melt pool, plume, and spatter tracking in laser powder bed fusion using DBSCAN. In Christopher Niezrecki and Saman Farhangdoust, editors, *Digital Twins, AI, and NDE for Industry Applications and Energy Systems 2025*, page 21. SPIE, May 2025. doi:10.1117/12.3051110

Camera Module Design for Inter-layer LPBF Monitoring

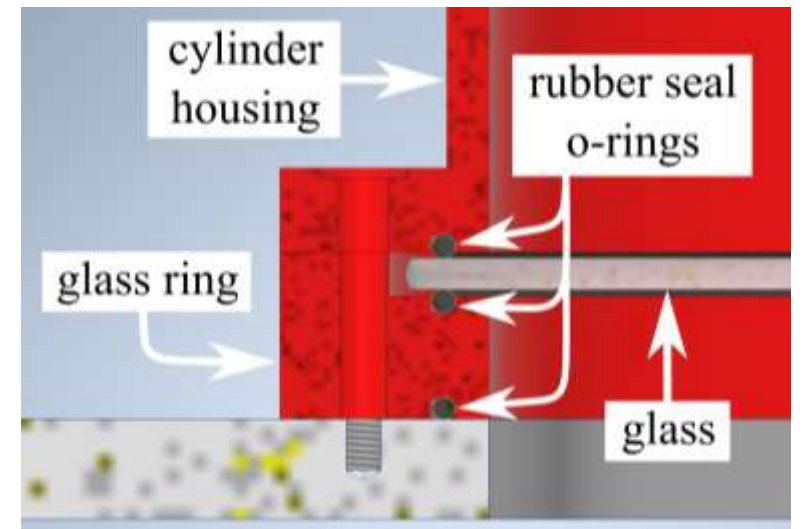
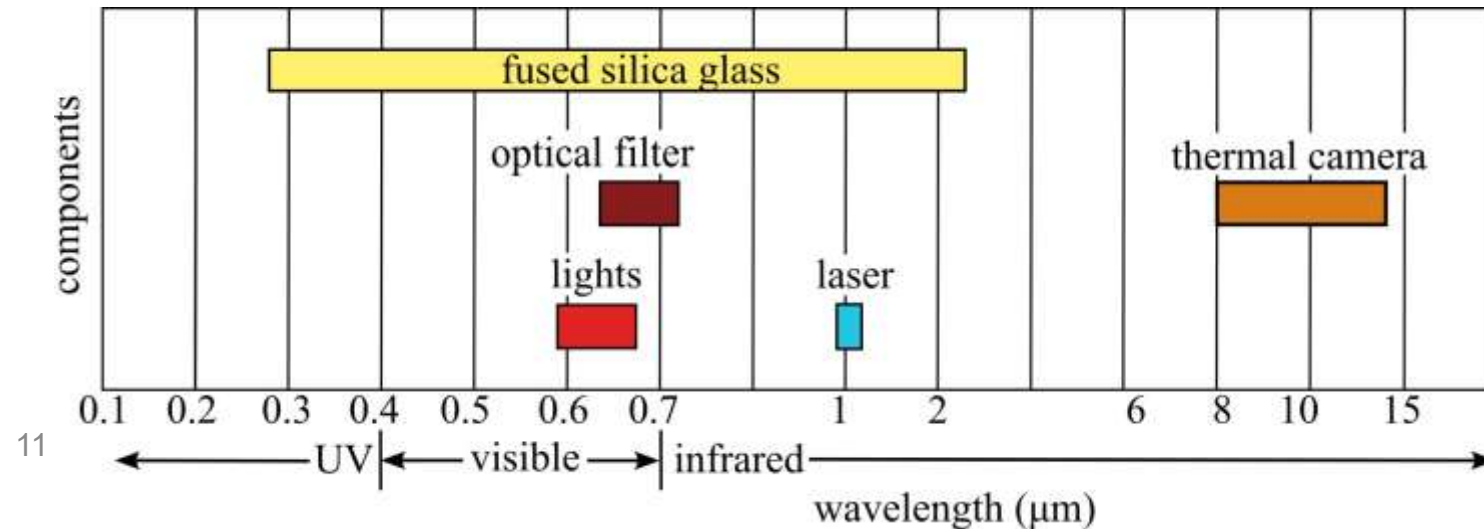
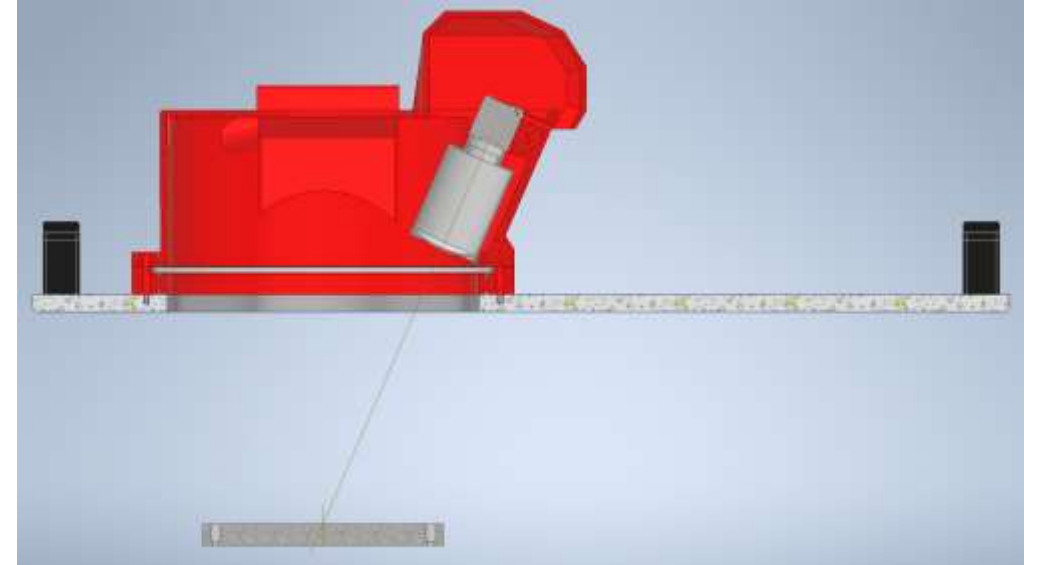
Optical Camera System

- FLIR Blackfly camera, 20 MP resolution
- USB 3.0 interface, 1" monochrome sensor
- 35 mm/F1.8 Edmund Optics lens
- 660 nm bandpass filter, red LED illumination at 625 nm



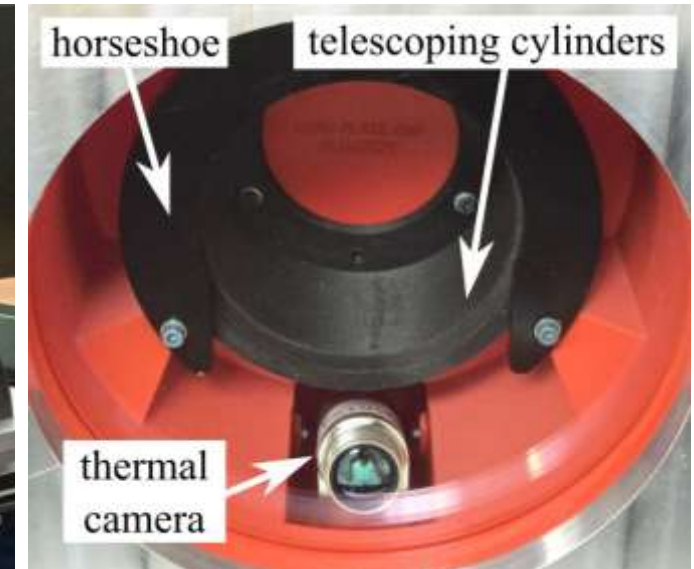
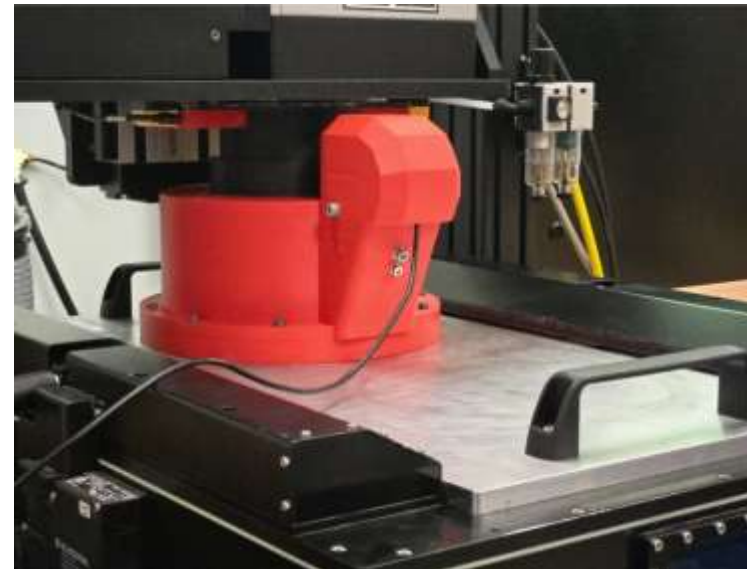
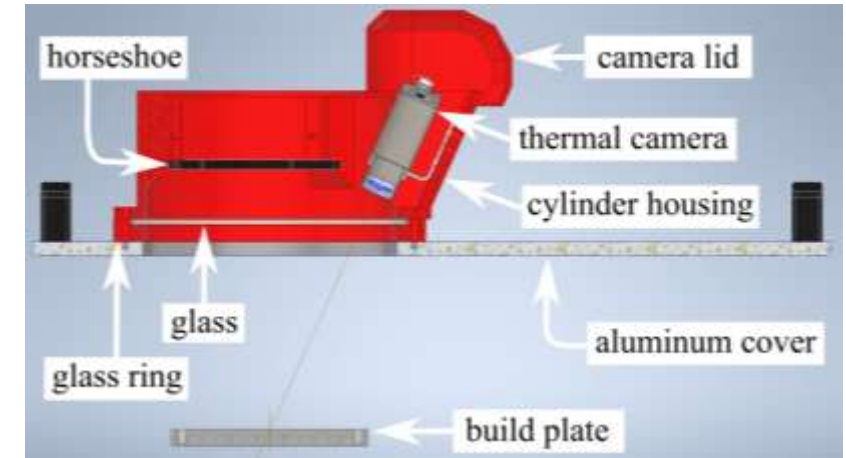
Sealed Chamber Design

- Stable argon gas environment for printing
- Chamber compatible with vacuum conditions if needed
- Sealing achieved with greased rubber o-rings



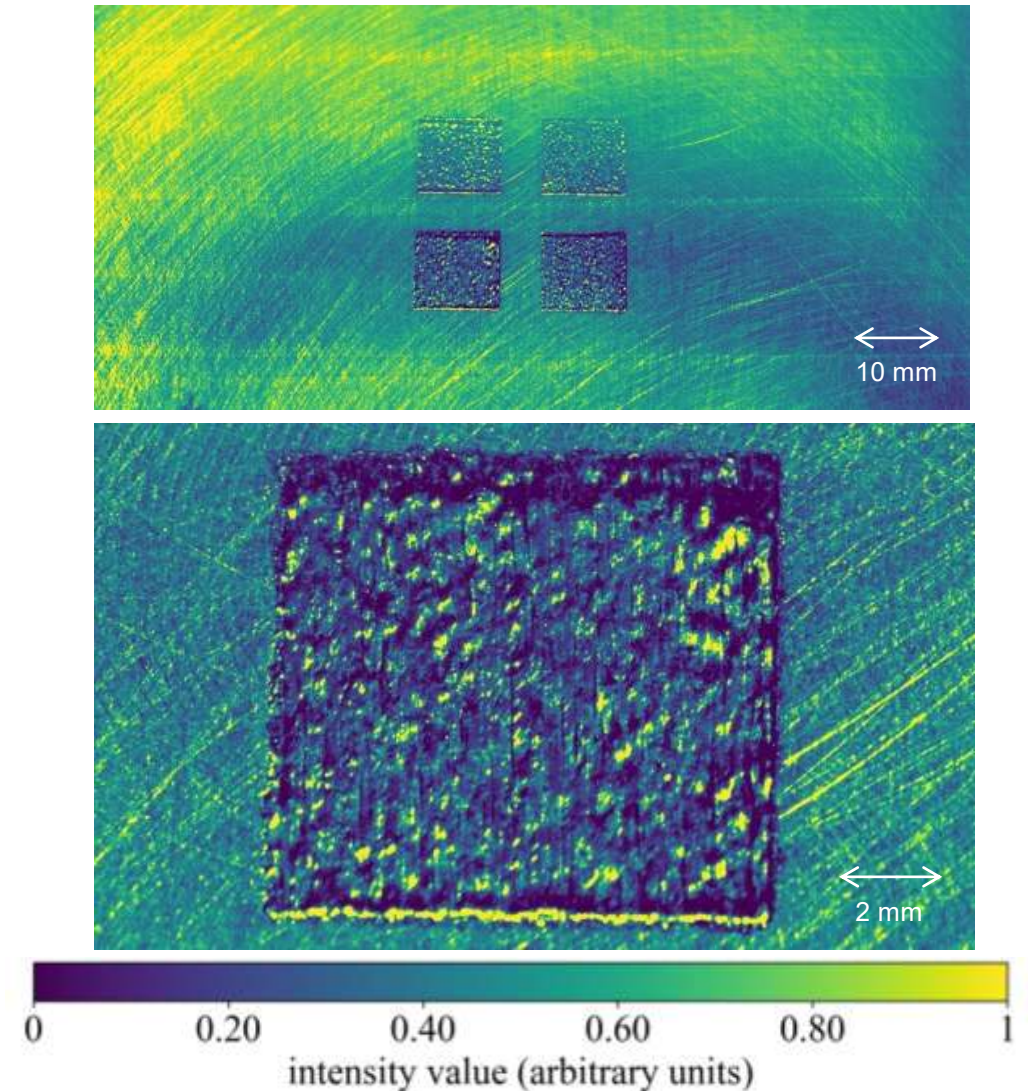
Thermal Camera System

- Optris Xi 400, 382 × 288 pixels
- 80 fps video, 18° × 14° field of view
- Digital focal depth control, USB communication via PIX Connect software



Inter-Layer Optical Imaging

- Captures the build surface after each layer is deposited
- Observes melt tracks, surface morphology, and spatter distribution
- Monochrome imaging enhances contrast of surface features
- Enables layer-by-layer assessment of build quality



Utilizing Thermal Imaging for Real-time Defect Detection

Battery Tab Welding

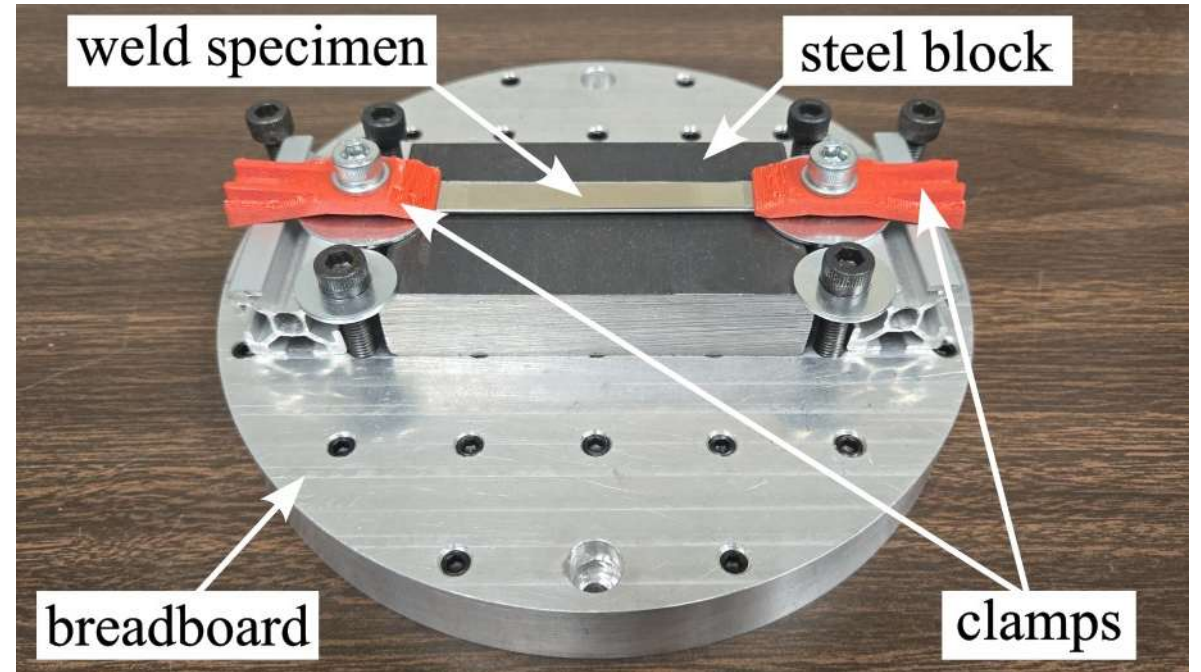
- Traditional battery pack manufacturing follows a cell-to-module-to-pack architecture.
- This approach is now shifting toward direct cell-to-pack integration.
- The long-term objective is to increase weld speed without compromising quality, enabling higher production efficiency and greater profitability.



Cell to Pack manufacturing method that will require high-quality, high-throughput battery tab welding on assembly lines.

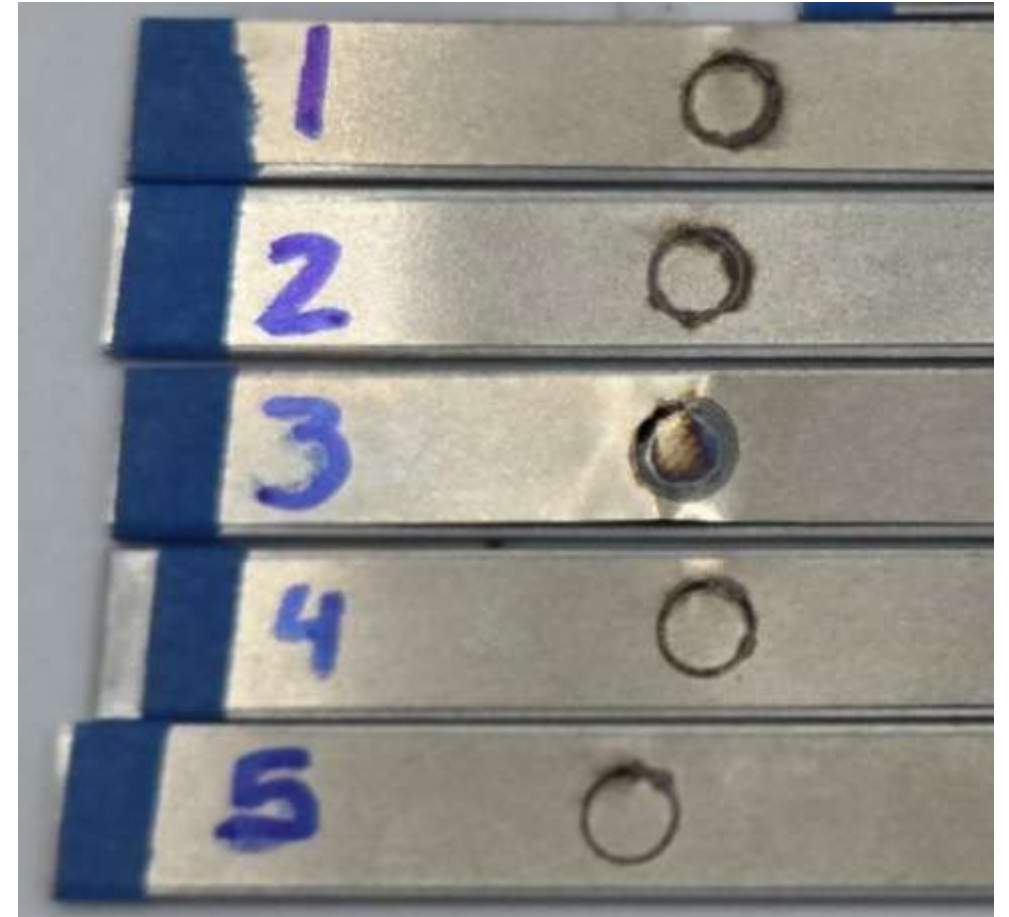
Welding Fixture

- An aluminum breadboard was used to construct a jig to hold each specimen.
- Specimens are held in clamps made from ABS-GF plastic.
- Alignment of specimen is done using an onboard laser.



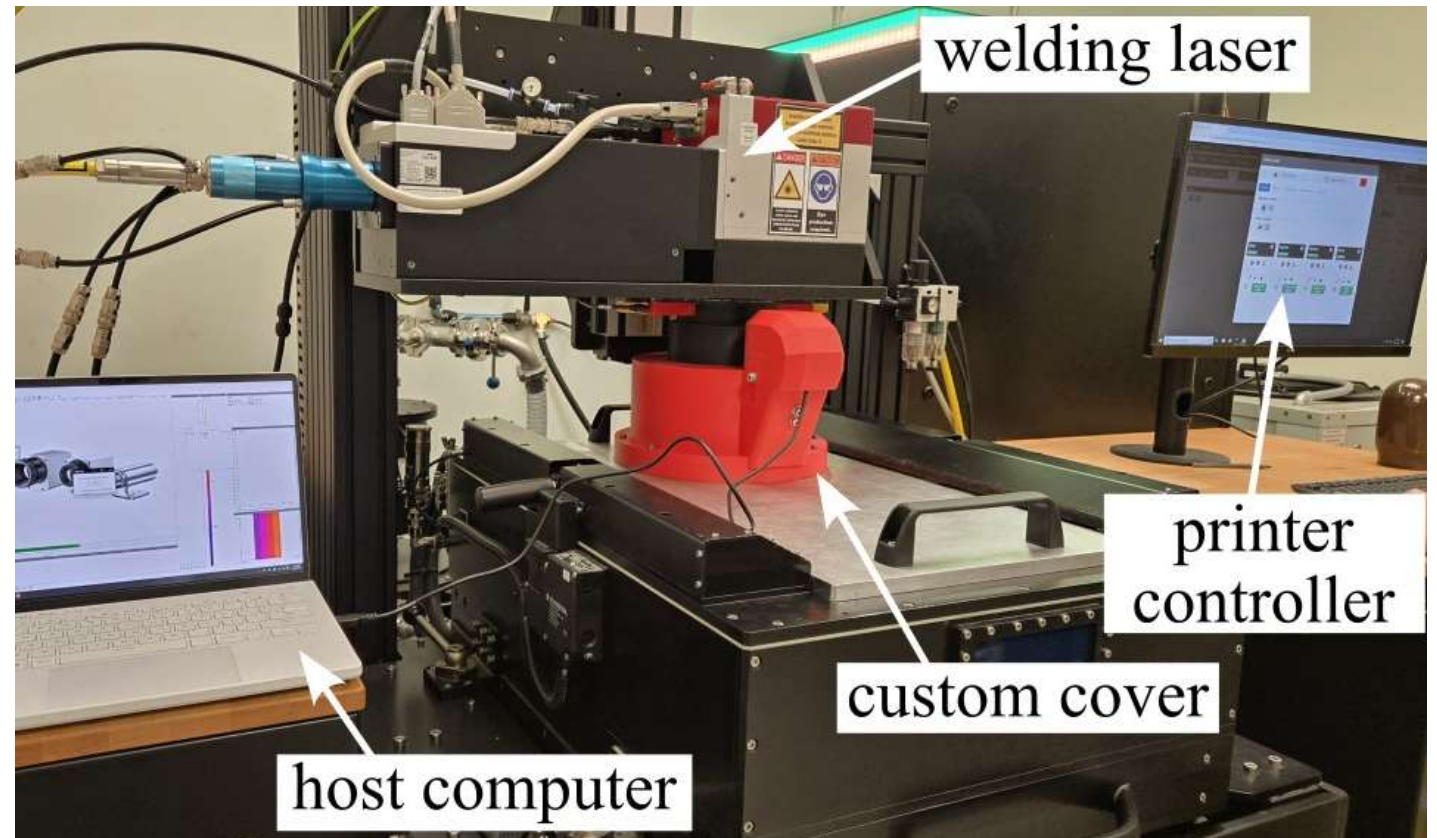
Specimens

- Dissimilar metal weld samples designed to replicate battery tab-to-case joints
- Comprised of:
 - Nickel-plated steel strip (tab material)
 - Stainless steel strip (battery casing surrogate)
- Dimensions:
 - Nickel strip: $\sim 0.15 \times 8 \times 100$ mm
 - Stainless steel: $\sim 0.6 \times 9.65 \times 85$ mm
- Materials chosen to mimic real cylindrical Li-ion battery configurations (e.g., 18650 cells)
- Each specimen was welded using a (~ 5 mm diameter circular laser pattern



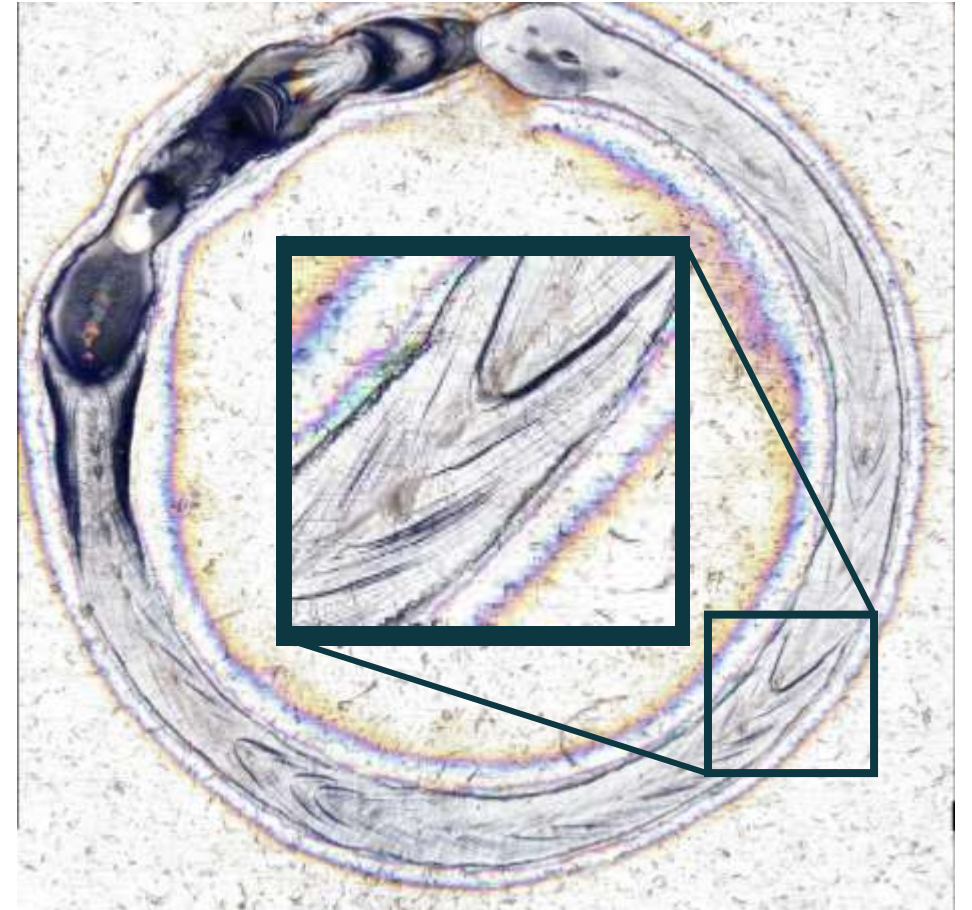
Test Setup

- Utilizing the custom cover equipped with a thermal camera
- No argon used for all experiment runs
- Anti-reflective glass removed to allow collection of accurate thermal data



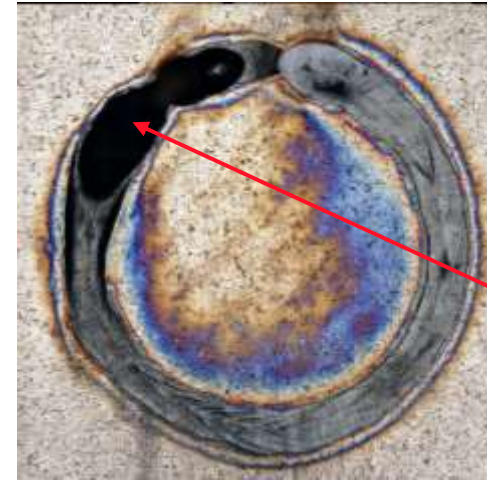
Laser Welding Parameters

- Laser power held constant at 300 W across all tests
- Welding speed varied (45 → 105 mm/s) to control heat input
- Constant spot size (150 μm) maintained for consistency
- Each condition was tested with replicate samples for comparison

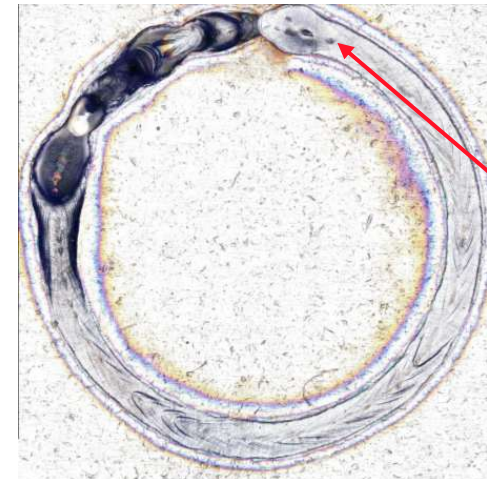


Defects in Battery Tab Manufacturing

- Process instabilities can produce weld defects
- Common defects include poor fusion and burn-through
- Defects can cause weak joints, increased resistance, and thermal runaway.
- Detection during welding is difficult.



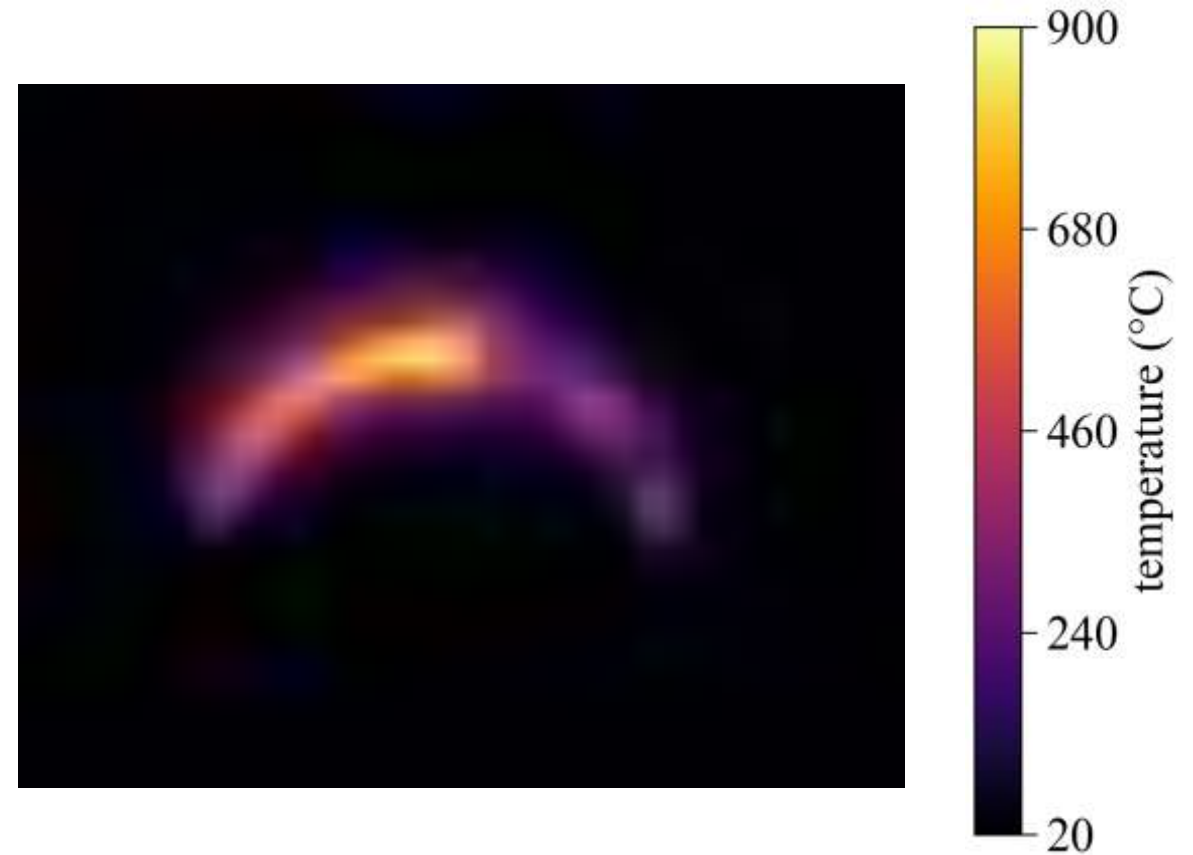
burn through



poor fusion

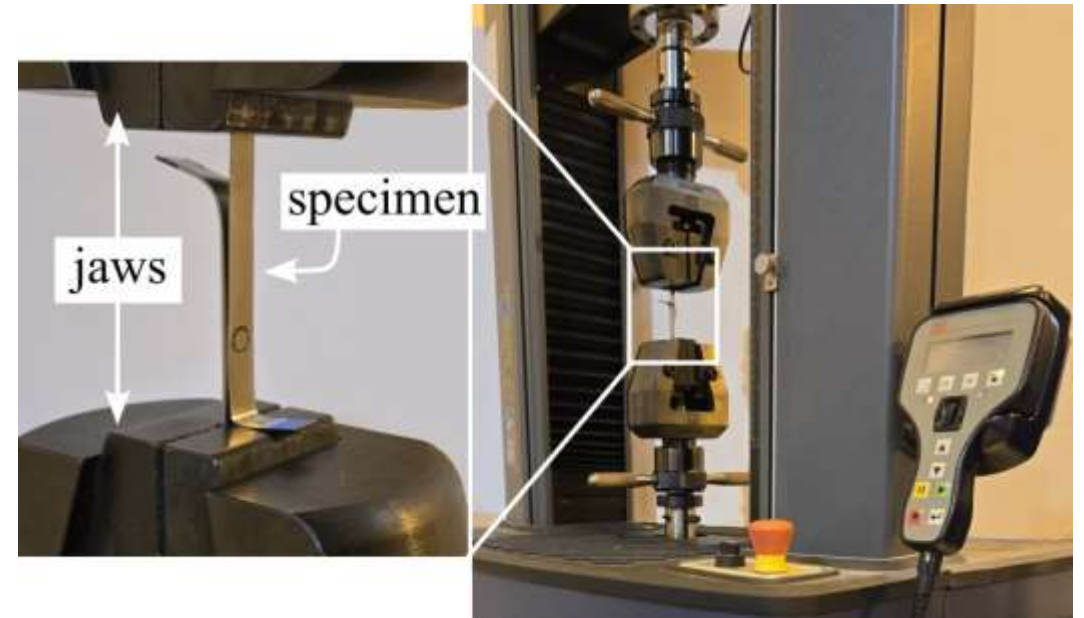
Thermal Imaging

- Thermal images show relative energy distribution in welds
- Absolute temperature is quantitative due to removal of window
- Identifies melt pool stability and cooling behavior
- Used to detect anomalies during the build



Tensile Testing

- Purpose: test weld strength
- Tensile test is qualitative, observing failure mode and location
- Quantitative tensile data not used for this project.



Test	Failure location
1	nickel strip
2	nickel strip
3	n/a
4	nickel strip
5	nickel strip
6	weld
7	nickel strip
8	nickel strip
9	n/a
10	both

Tensile Data

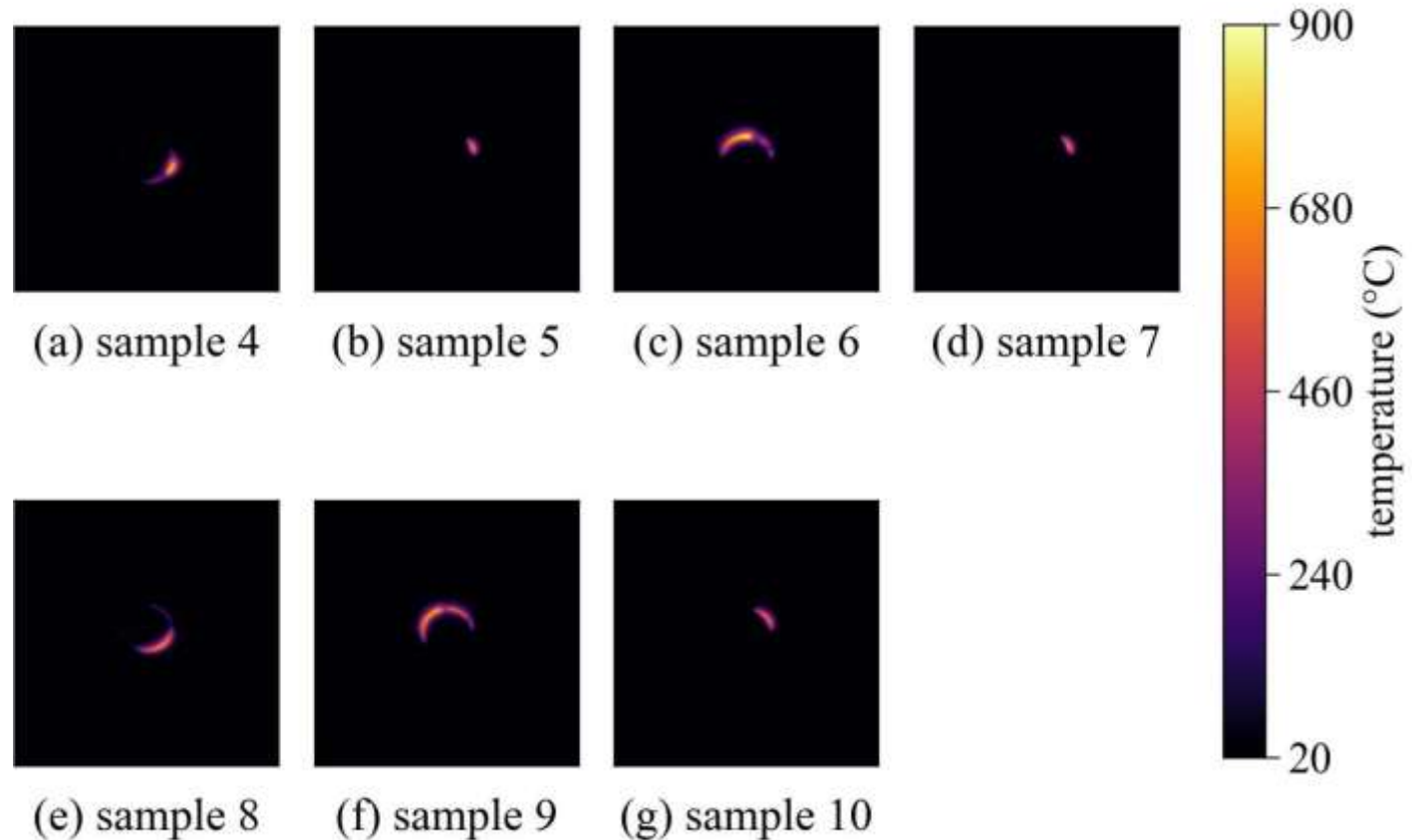
- Weld results: Around of 70% success rate among samples. Initial quality testing was done using a shim.
- Specimen 16 was not considered for this experiment
- Weld 6 was partially welded but still considered failed.



1	Passed
2	Passed
3	Failed
4	Passed
5	Passed
6	Failed
7	Passed
8	Passed
9	Failed
10	Passed

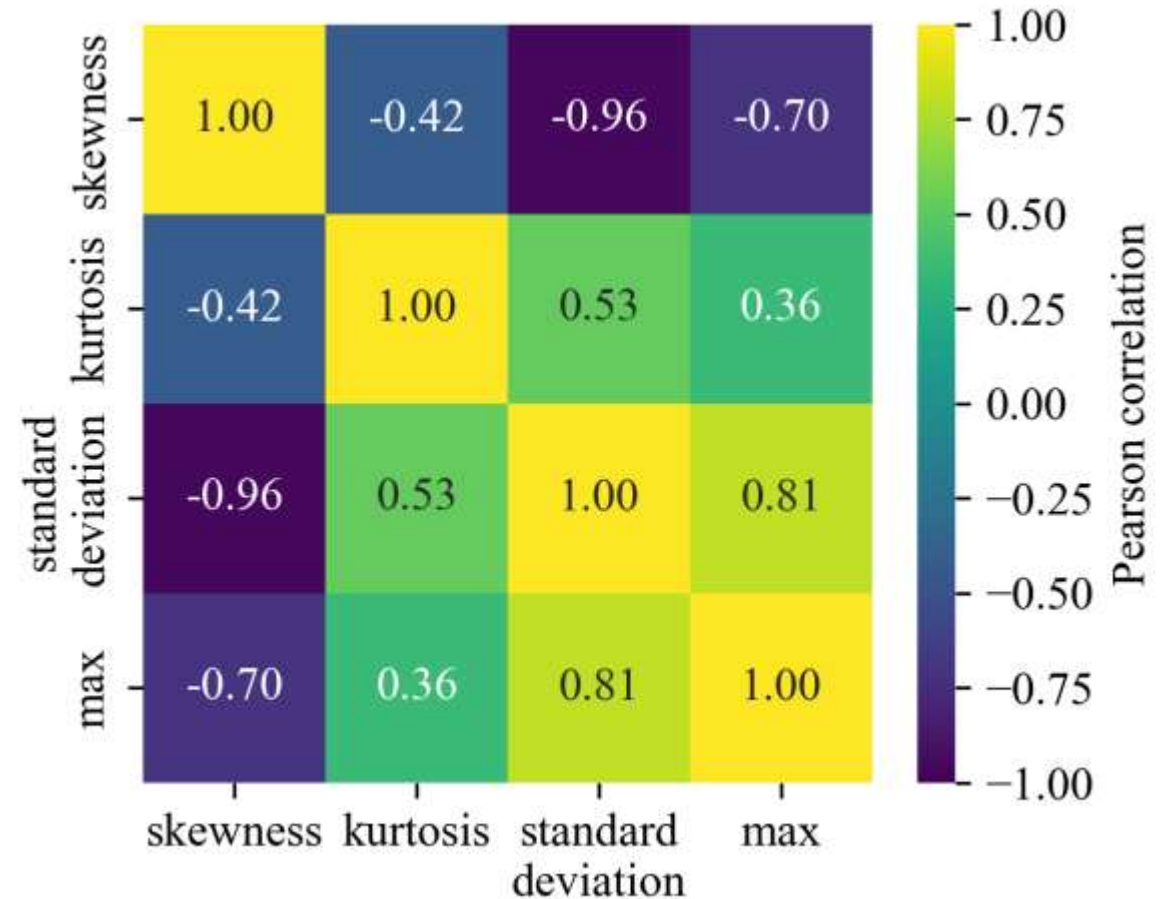
Feature Extraction from Thermal Images

- Region of interest (ROI) isolates the weld zone for consistent analysis
- Extracted statistical features from each image:
 - Maximum temperature (peak intensity)
 - Standard deviation (heat spread/variability)
 - Skewness (asymmetry of heat distribution)
 - Kurtosis (concentration of heat)



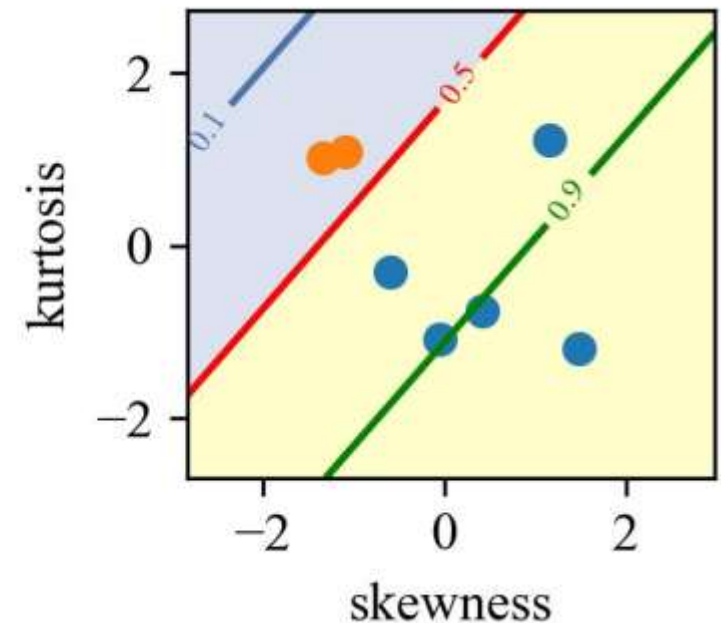
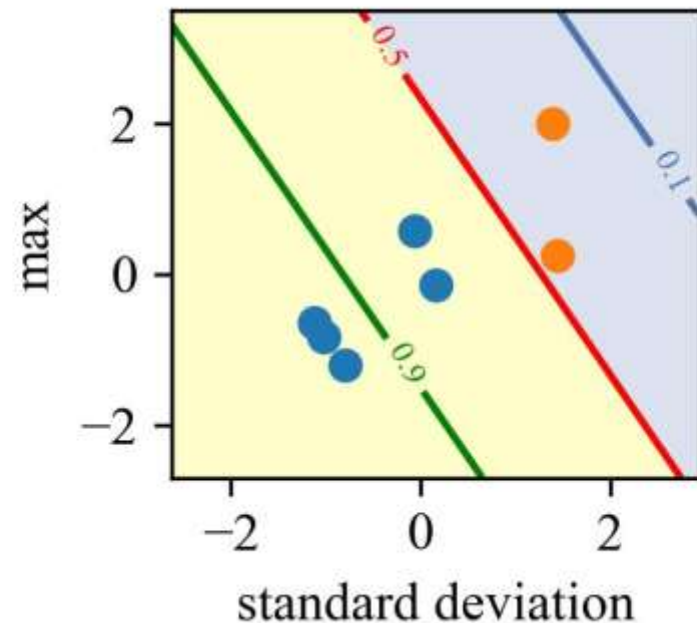
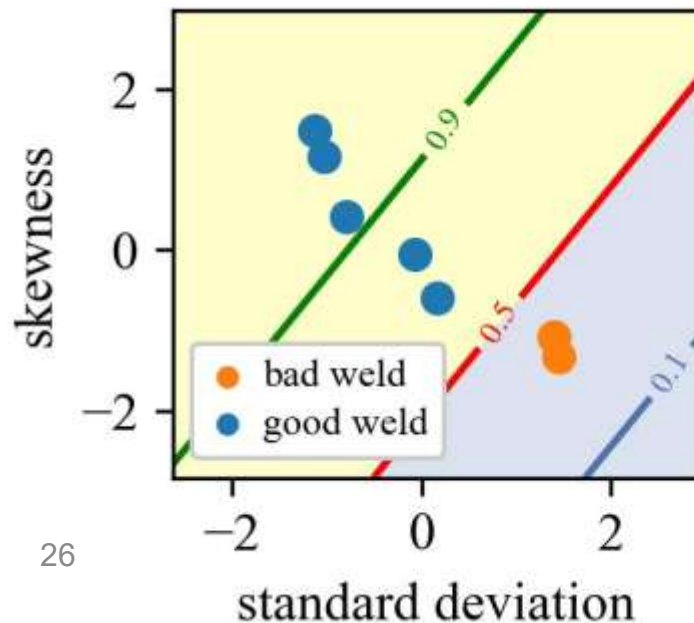
Feature Correlation Analysis

- Correlation matrix shows relationships between selected thermal features
- Strong correlations (e.g., std. dev. vs. max) indicate overlapping information about weld behavior
- Weak correlations highlight complementary features that add unique insight for classification



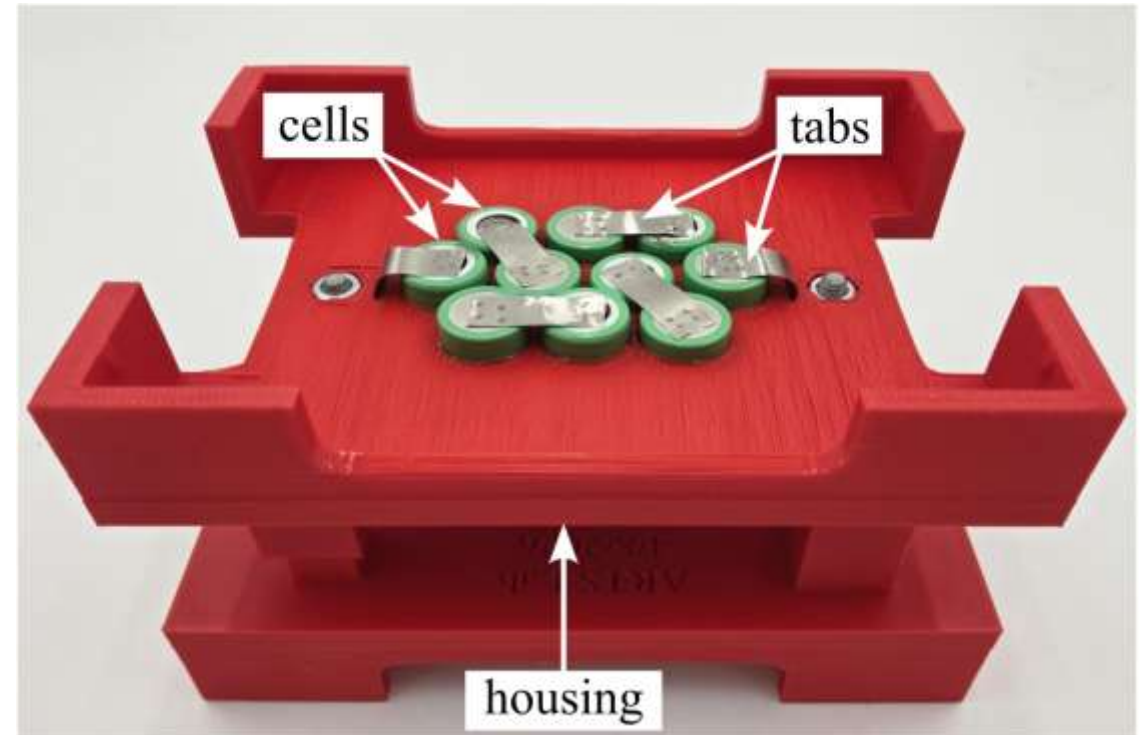
Weld Quality Classification

- Logistic regression applied to thermal features to separate good vs. bad welds using decision boundaries
- Feature pairs (e.g., std. dev., skewness, kurtosis, max temp) show clear class separation and predictive capability



Future Work

- Expand data processing to include both profilometer and quantitative tensile data.
- Develop a “large” dataset of 100 laser battery tabs welds, post-process inspection, and destructive testing.
- Increase the number of samples to improve weld quality classification.



The Open-source Smart Pack for Advanced Research and Control (SPARC) is being developed at USC.

<https://github.com/ARTS-Laboratory/Smart-Pack-for-Advanced-Research-and-Control>

ACKNOWLEDGEMENT

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Discussion

In Situ Monitoring of Powder Bed Fusion Additive Manufacturing

<https://github.com/ARTS-Laboratory/In-situ-monitoring-of-powder-bed-fusion-additive-manufacturing>



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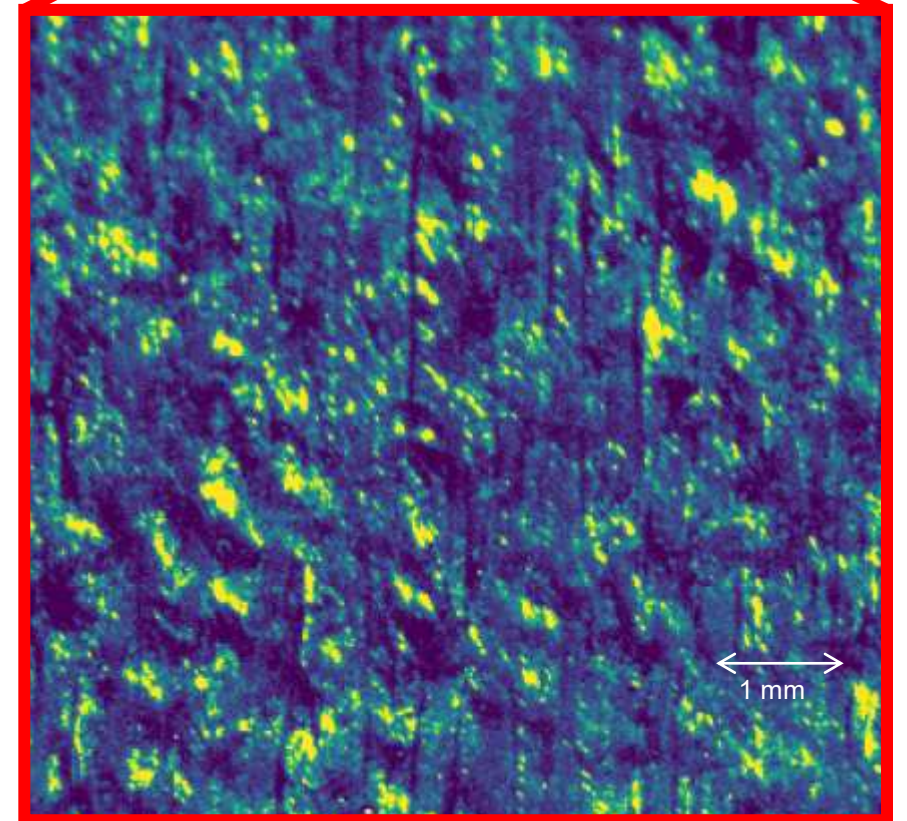
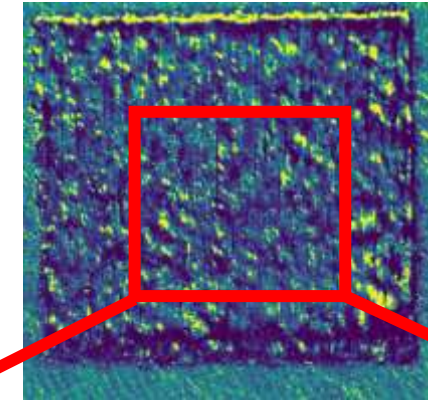
Profilometer Data

- Purpose: Collecting surface roughness data for each sample.
- Images are processed to find height of each pixel
- Data not used for this project



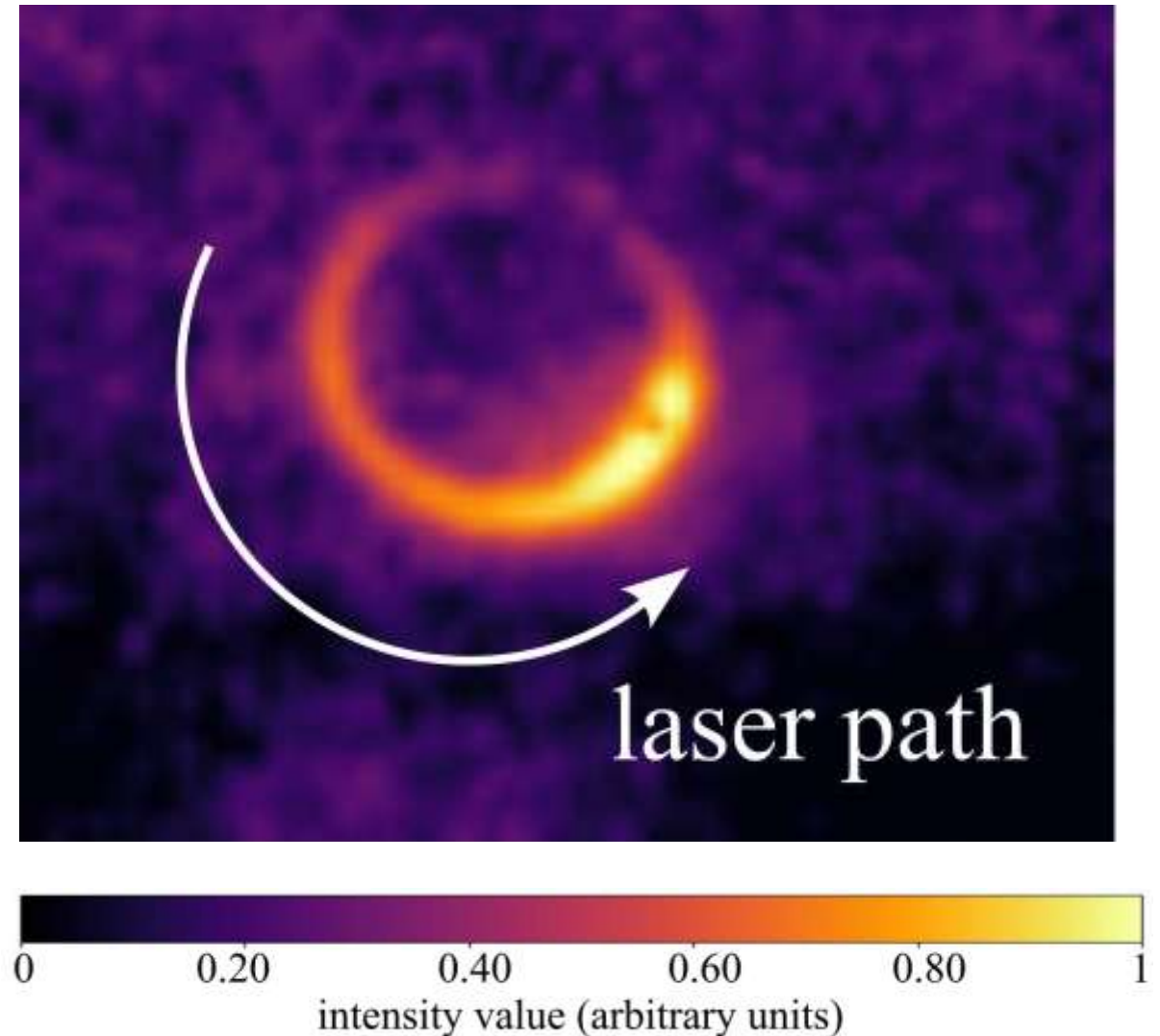
Optical Imaging Resolution

- Sample width: 10 mm field of view
- Spatial resolution: ~50 pixels per mm (20 μm per pixel)
- Individual melt tracks ~15 pixels wide (~300 μm)
- Resolution sufficient to capture fine surface and track features



Thermal Imaging Insights

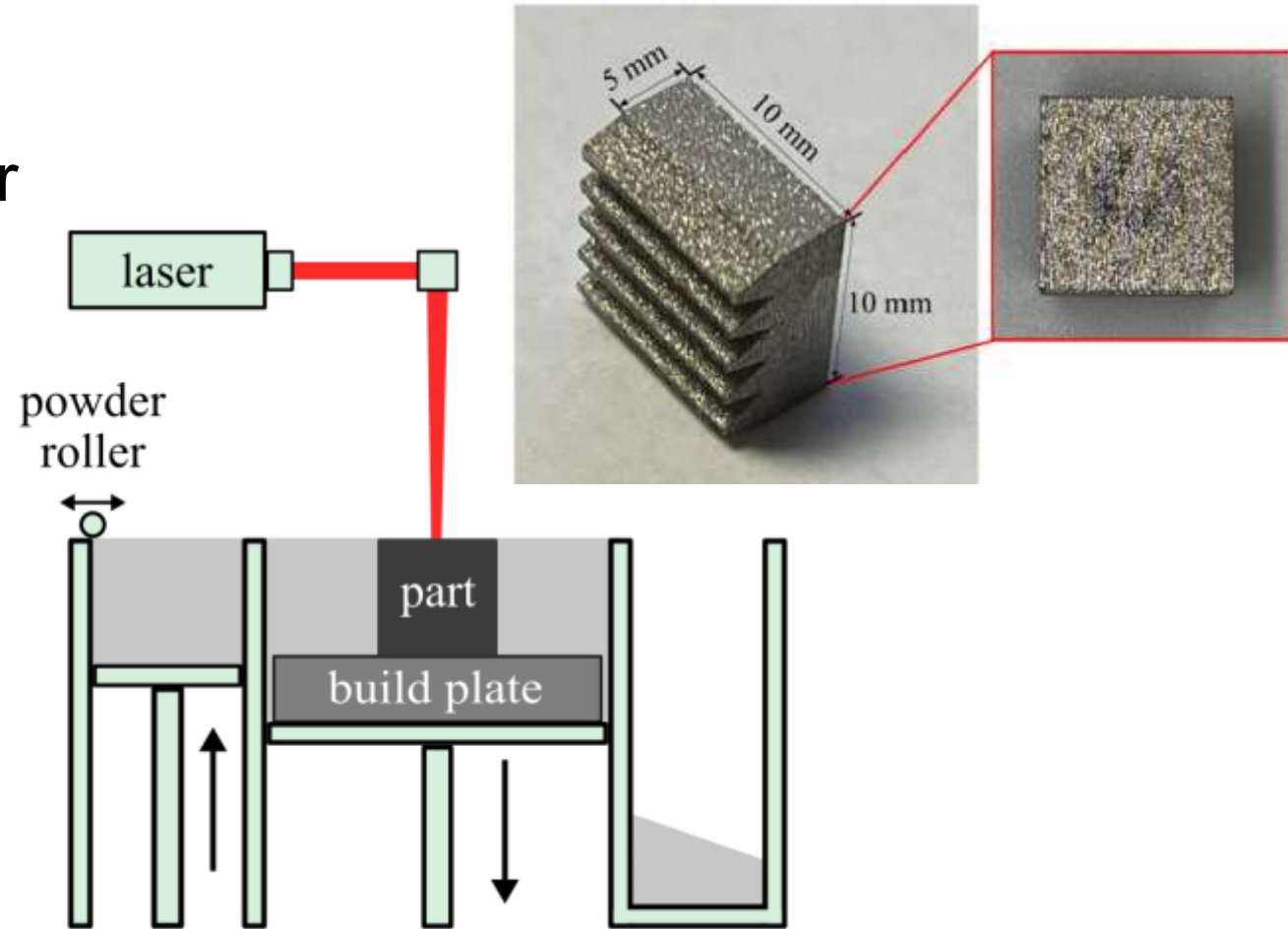
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Surface Quality and Melt Pool Behavior

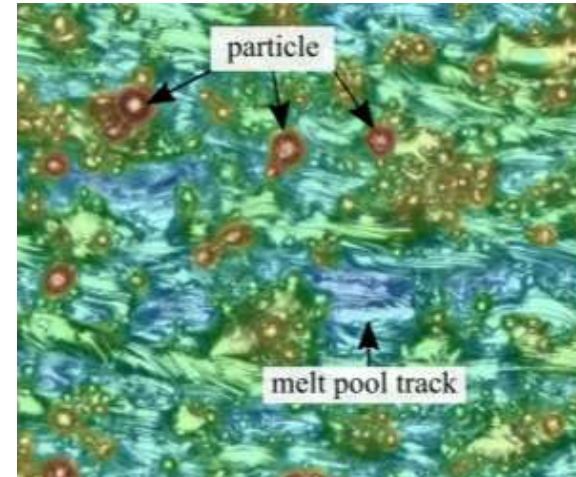
Laser Powder Bed Fusion (LPBF)

- Metal additive manufacturing process using a scanning laser
- Powder is melted layer-by-layer to build complex geometries
- Enables high precision and material efficiency
- Widely used in aerospace, biomedical, and automotive industries



Defects in LPBF Manufacturing

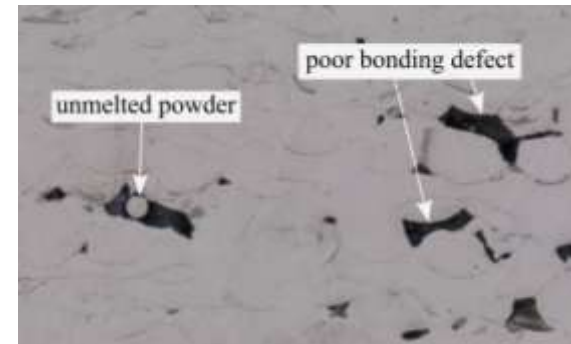
- Process instabilities can introduce internal and surface defects
- Common defects include porosity, keyhole formation, and distortion
- Surface roughness and lack of fusion affect mechanical properties
- Detecting these defects during printing remains challenging



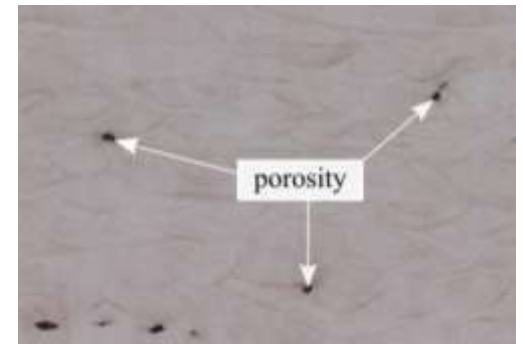
Surface roughness



Distortion



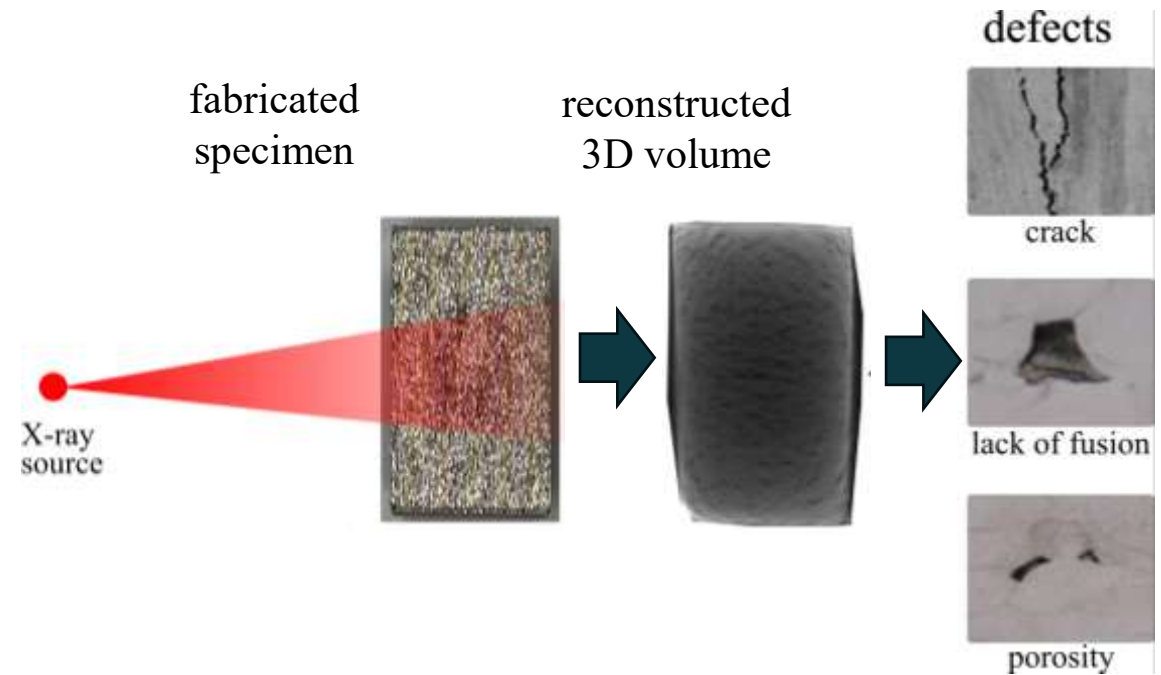
Keyhole



Porosity

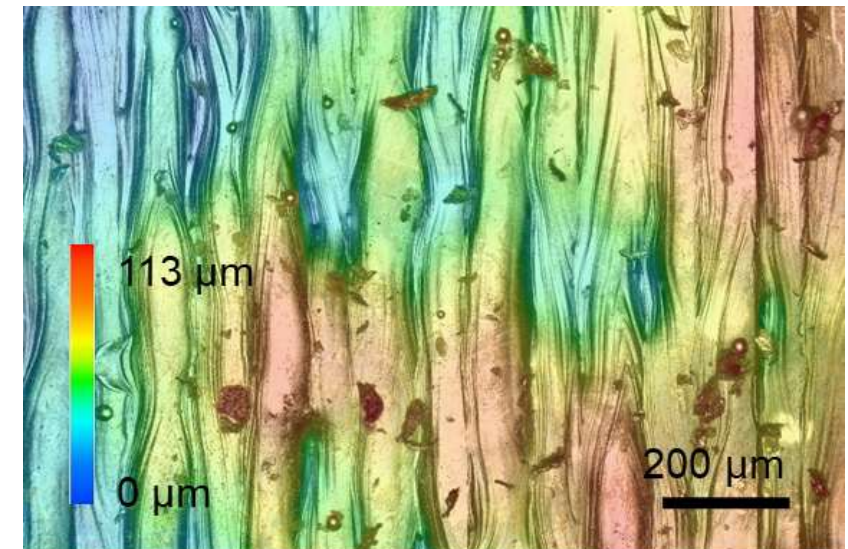
Challenges in LPBF Monitoring

- Defects are typically identified after printing using XCT or microscopy
- Post-process inspection is time-consuming and expensive
- Limited ability to detect defects during fabrication
- Real-time monitoring is needed for improved process control



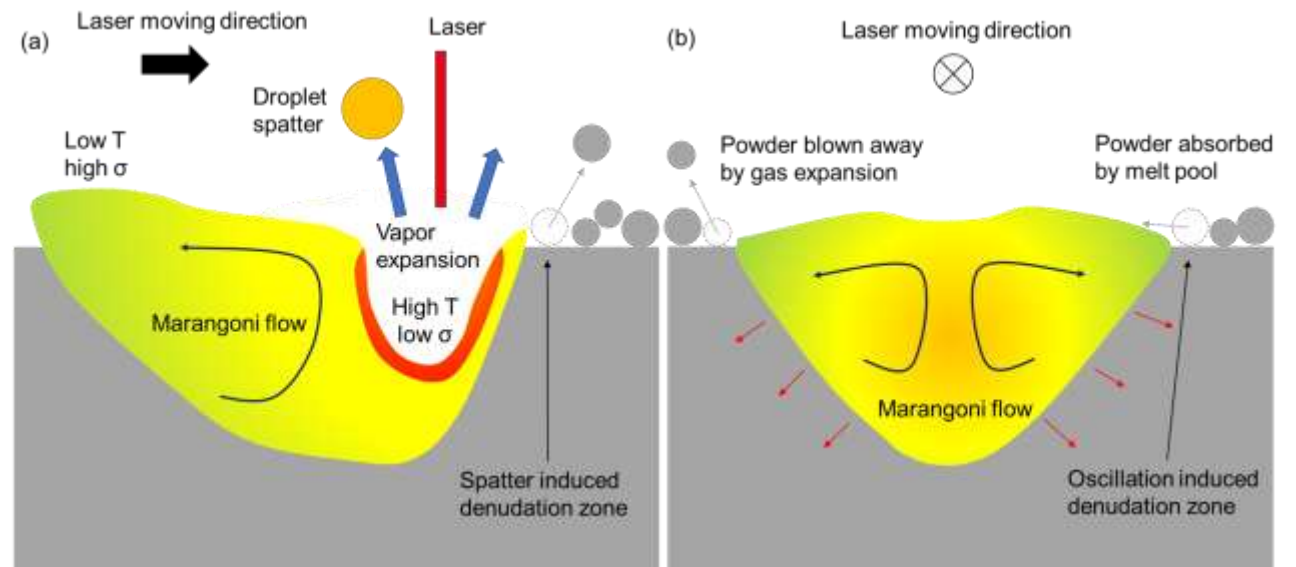
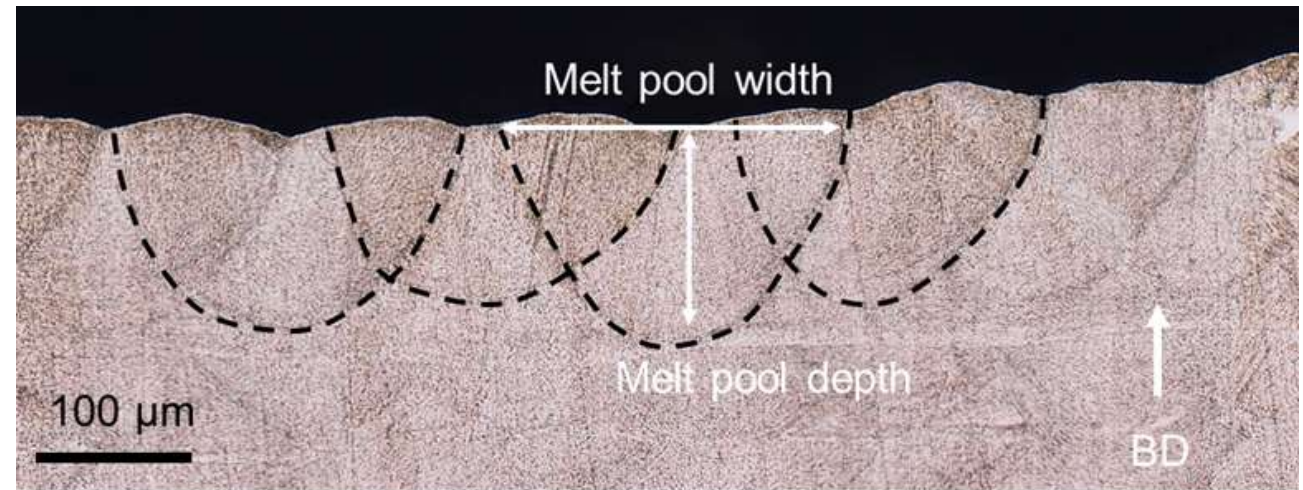
Surface Roughness

- Surface roughness is a key indicator of part quality and performance
- Strongly influenced by melt pool dynamics and spatter formation
- Low energy input can cause discontinuities and incomplete melting
- Excessive energy can create humps and unstable melt behavior



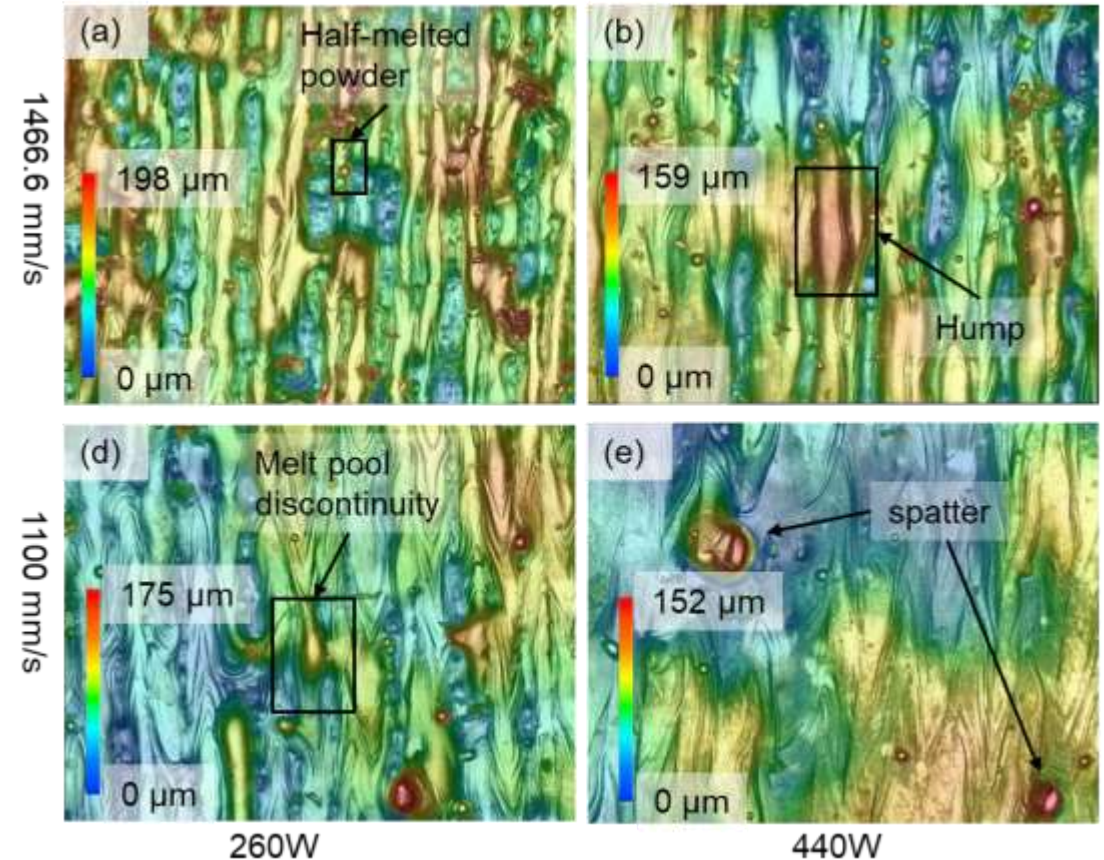
Melt Pool Stability and Spatter

- Melt pool behavior directly affects layer quality
- Stable melt pools produce smooth and continuous tracks
- Instability leads to spatter ejection and irregular solidification
- These effects increase surface roughness and defect probability



Impact of Laser Power on Surface Quality

- Low laser power leads to melt pool discontinuities and track gaps
- Moderate power improves melt pool continuity and surface finish
- High power can produce spatter and hump formation
- Optimizing laser power is critical for consistent surface quality



Eliminating melt pool discontinuities, valleys, and spatter improves the top surface quality.

LPBF Printing Chamber Environment

- Printing occurs in a controlled atmosphere chamber
- Inert gases (e.g., argon) prevent oxidation of reactive materials
- Vacuum conditions are relevant for space-based welding
- Optical access must maintain sealing and laser safety
- Argon was not used during this experiment.

