



National Center for Additive Manufacturing Excellence

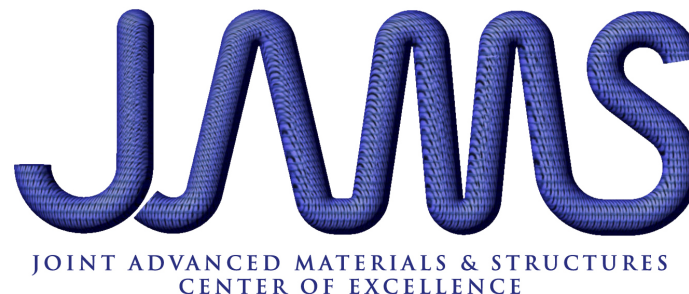
**Factors Affecting Qualification/Certification - Evaluating the Criticality of Inherent Anomalies/Defects on the Fatigue Behavior of Additively Manufactured Ti-6Al-4V Parts**

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Projects sponsored by: Federal Aviation Administration (FAA)

# Introduction

- **Project Title:** Factors Affecting Qualification/Certification - Evaluating the Criticality of Inherent Anomalies/Defects on the Fatigue Behavior of Additively Manufactured Ti-6Al-4V Parts
- **Principal Investigator:** Nima Shamsaei  
(See next slide for complete list of participants)
- **FAA Technical Monitor:** Kevin Stonaker
- **Source of matching contribution:** Faculty time and graduate research assistant tuition



# Project Team



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**NCAME Project Team**  
Auburn University

8 Senior Investigators  
10 Graduate Research Assistants

**PI:** Nima Shamsaei (Mechanical Engr.)

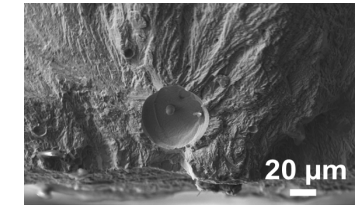
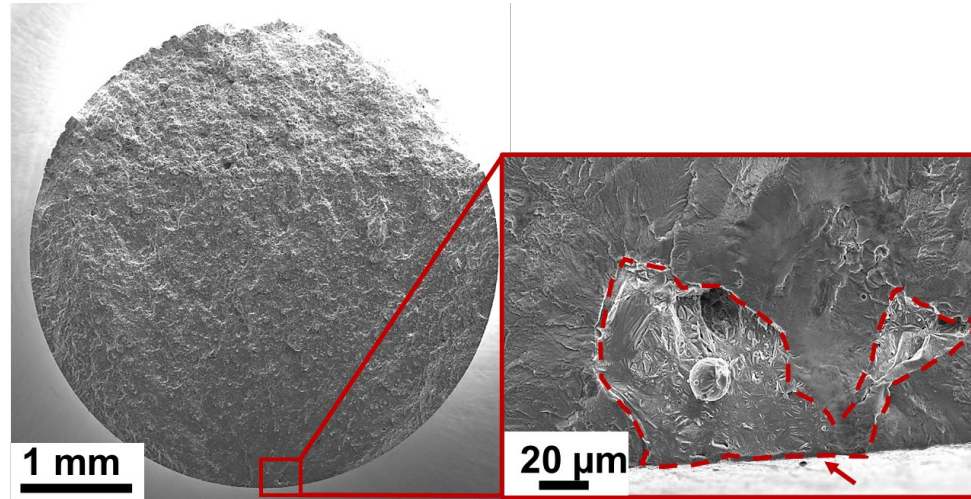
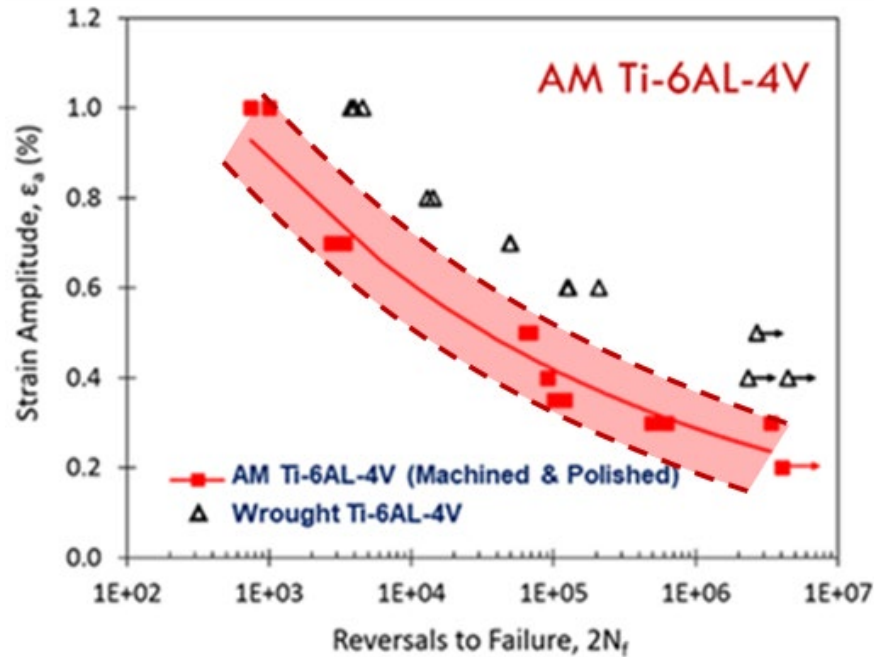
### Co-PIs:

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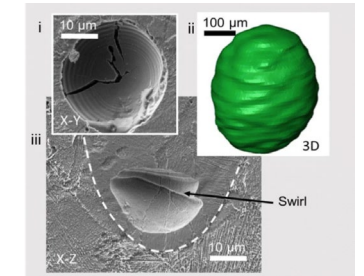
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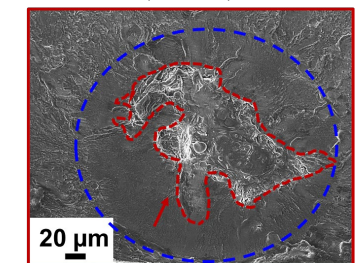
# Background



Gas-entrapped pores (GEPs)



Keyholes (KHs)



Lack of fusions (LoFs)

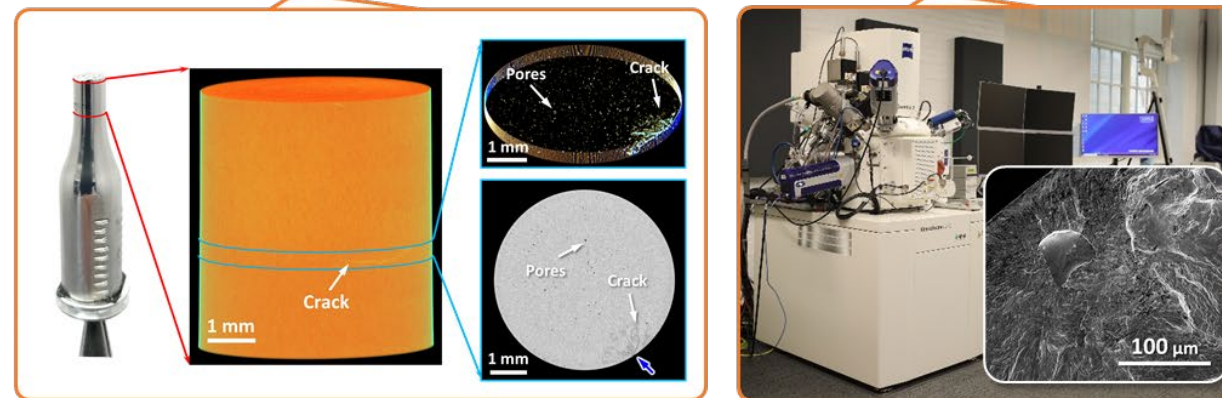
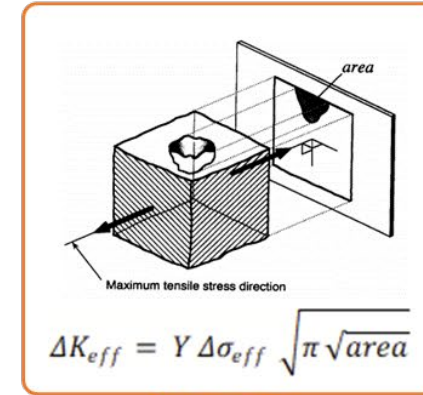
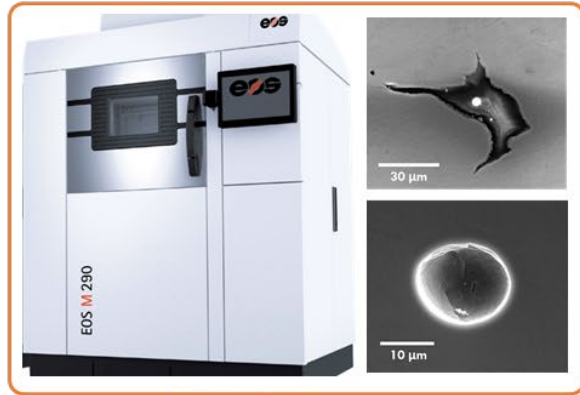
AM defects:

- Significantly reduce and introduce uncertainty to fatigue performance
- Pose great challenge for qualification/certification of AM parts

# Objective and Approach

- **Objective:** To quantify the detrimental effect of volumetric defects on mechanical properties of L-PBF Ti-6Al-4V Gr. 5
- **Approach:** Three steps are taken,
  - I. Explore process windows by varying laser power, scan speed, and hatching distance
  - II. Determine the criticality of volumetric defects on mechanical performance using specimens seeded with different defect types
  - III. Take advantage of machine learning and simulations wherever applicable

# Overall Scope

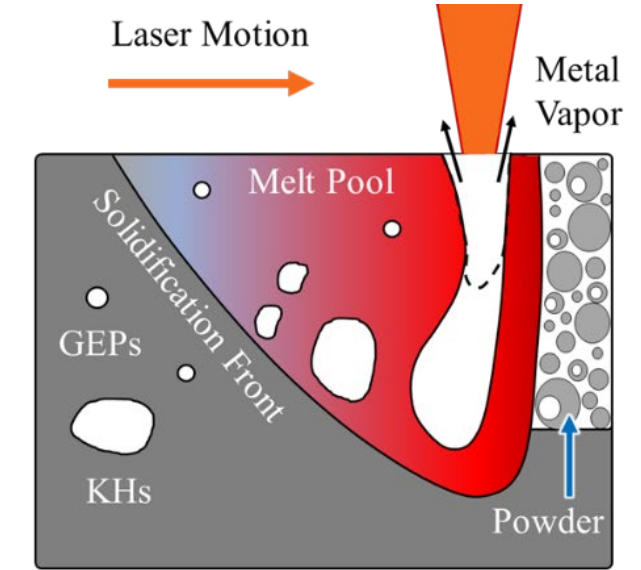
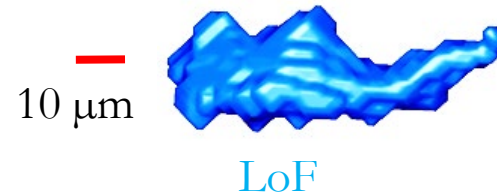
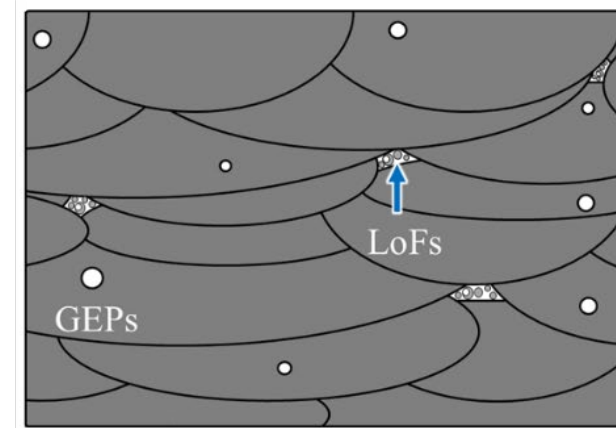
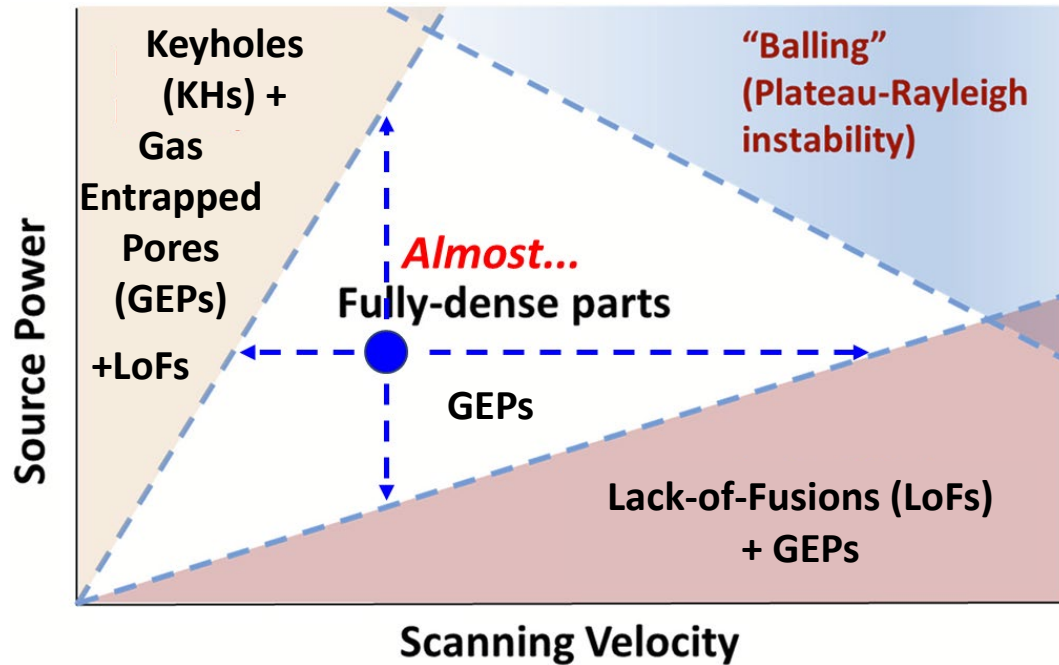


- AP&C Ti-6Al-4V Grade 5 powder (15-53 µm) was used as feedstock



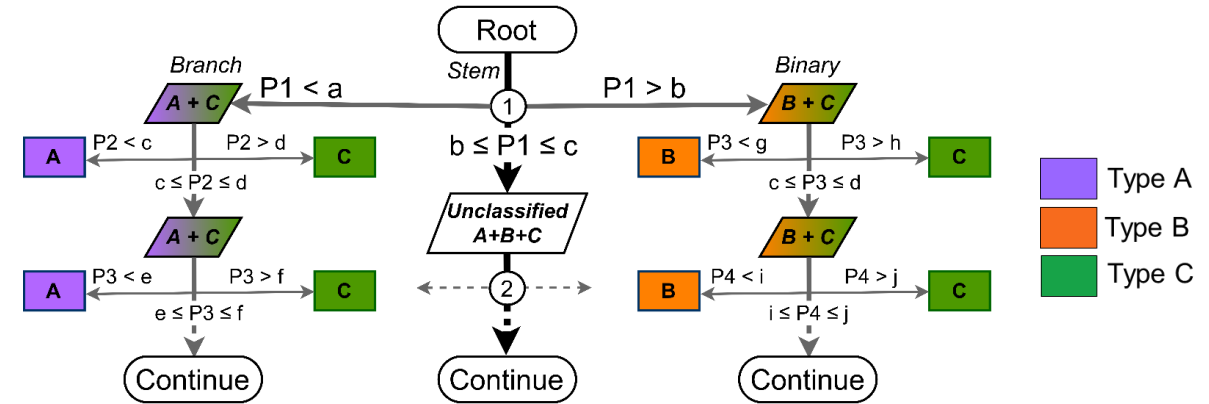
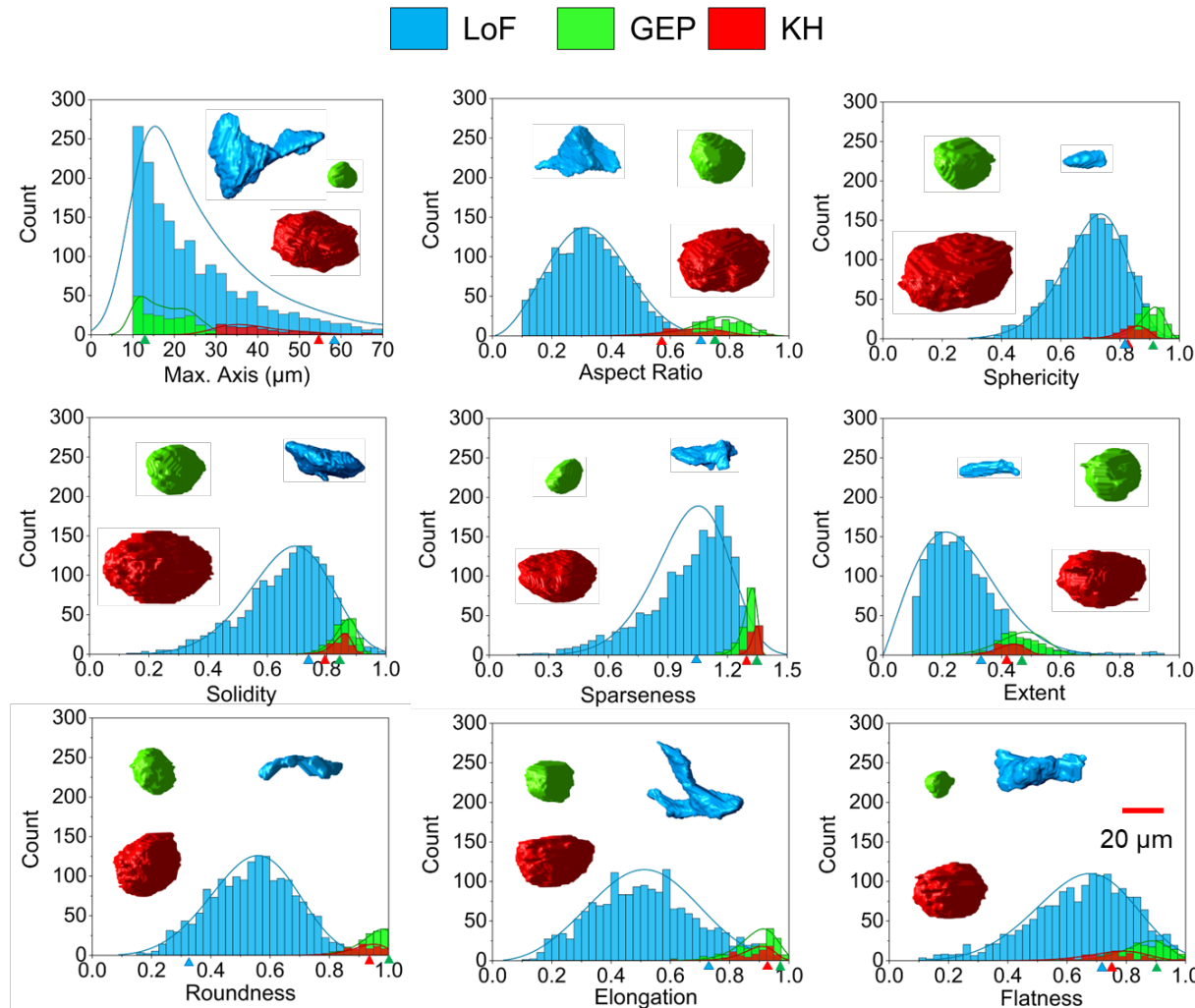
# Significance of Categorical Defect Feature

Schematic of the Process Window



- Categorical defects feature, i.e., defect types (KHs, GEPs, LoFs), is indicative of processing condition

# Decision Tree for Defect Classification

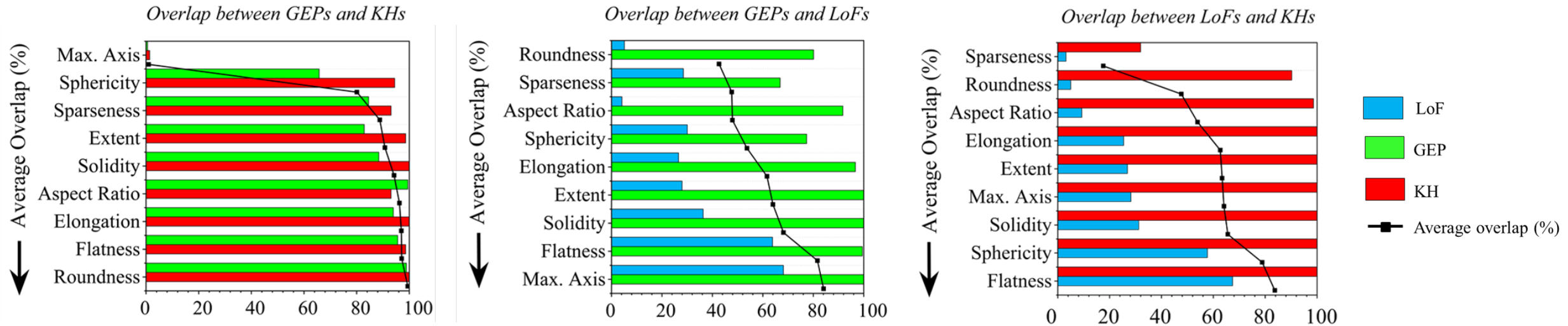


Note: Reported accuracy is only for Ti-6Al-4V Grade 5 material fabricated using EOS M290

- Morphological parameters such as max. axis, aspect ratio, sphericity, extent, solidity, sparseness, roundness, elongation, and flatness were used for defect classification



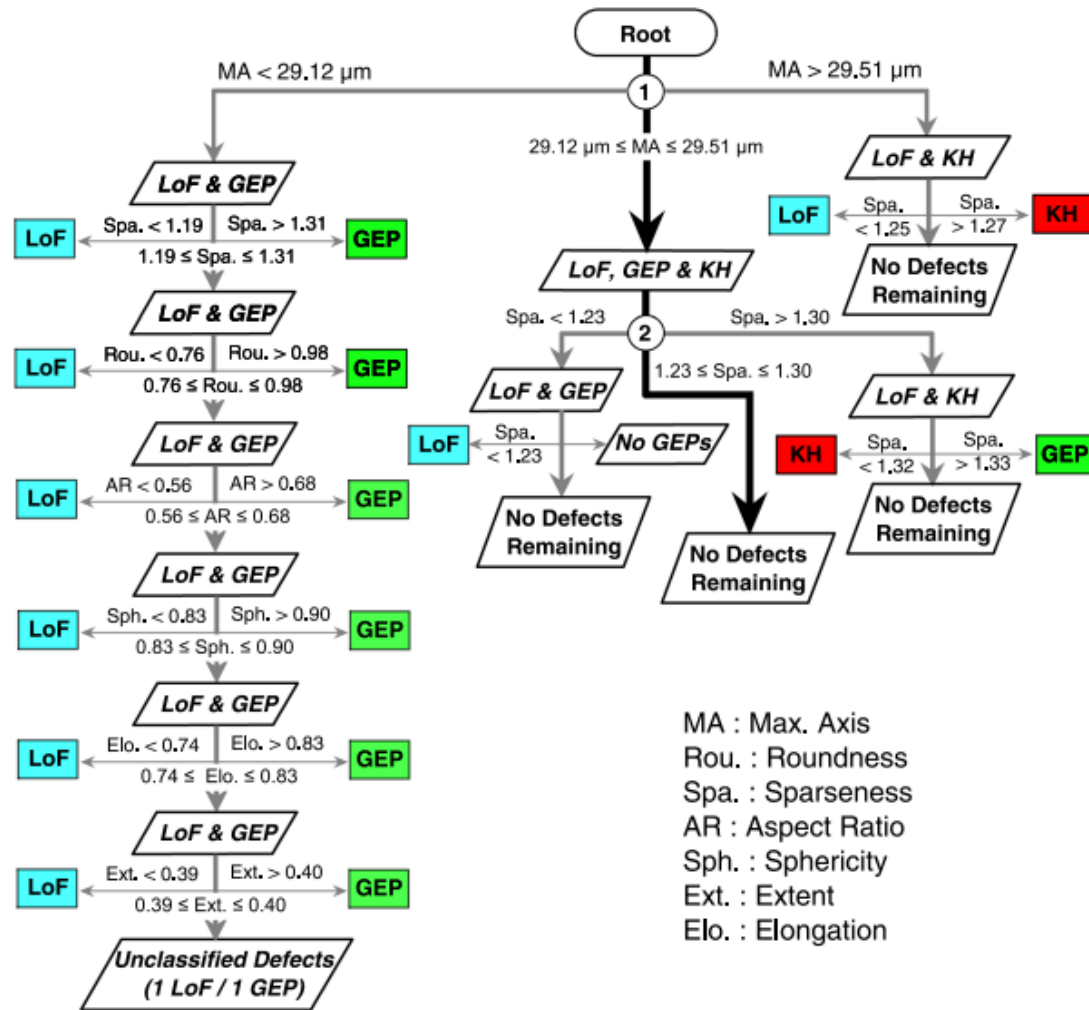
# Ranking the Most Discriminating Parameters



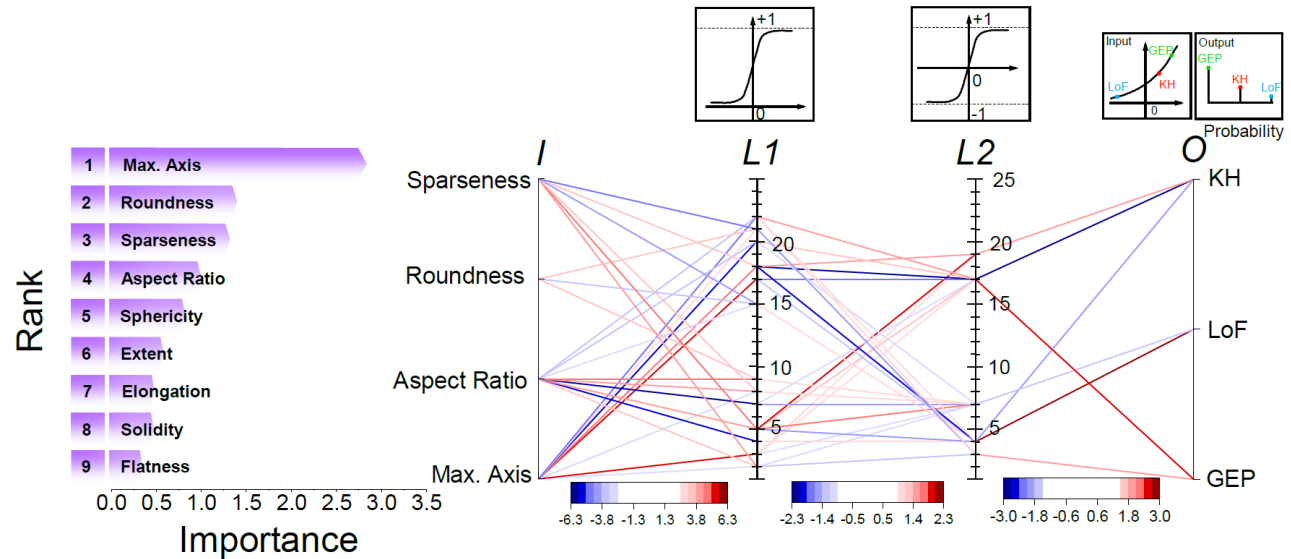
- The three defect types share overlaps of different degrees in the ranges of their morphological parameters; thus, employing only one or two parameters cannot uniquely determine a defect's type
- The discriminating potential of a morphological parameter depends on the pair of defect types:
  - Defect size (max. axis) best discriminates GEPs from KHs
  - Roundness and sparseness best discriminate GEPs and KHs from LoFs

# Defect Classification Methodology

## Decision Tree

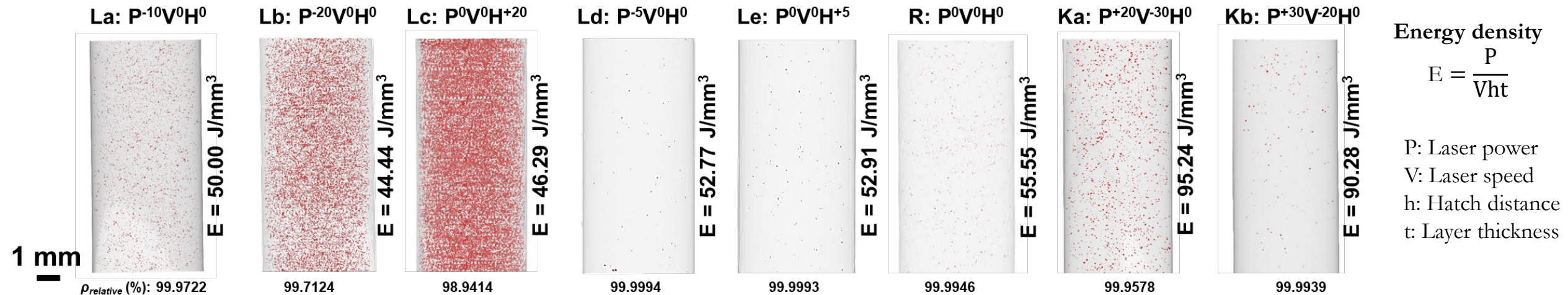


## Artificial Neural Network



- A defect classification methodology incorporating multiple morphological parameters had been developed
  - >98% accuracy when implemented into decision tree
  - >99% accuracy when implemented into artificial neural network

# Defect Contents: Fatigue Specimens

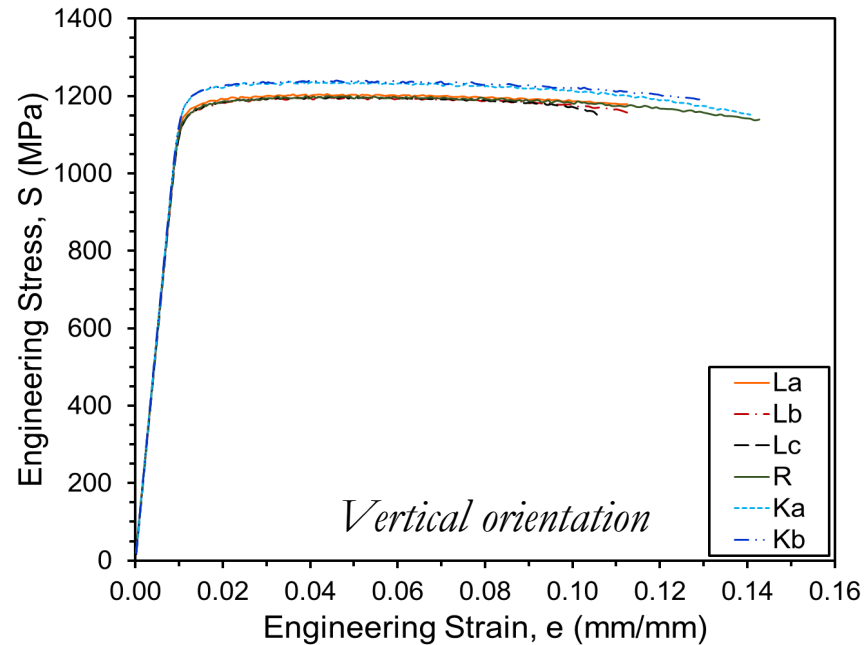


Note: X-ray computed tomography (XCT) was performed on vertical fatigue specimens with 5.5 μm voxel size

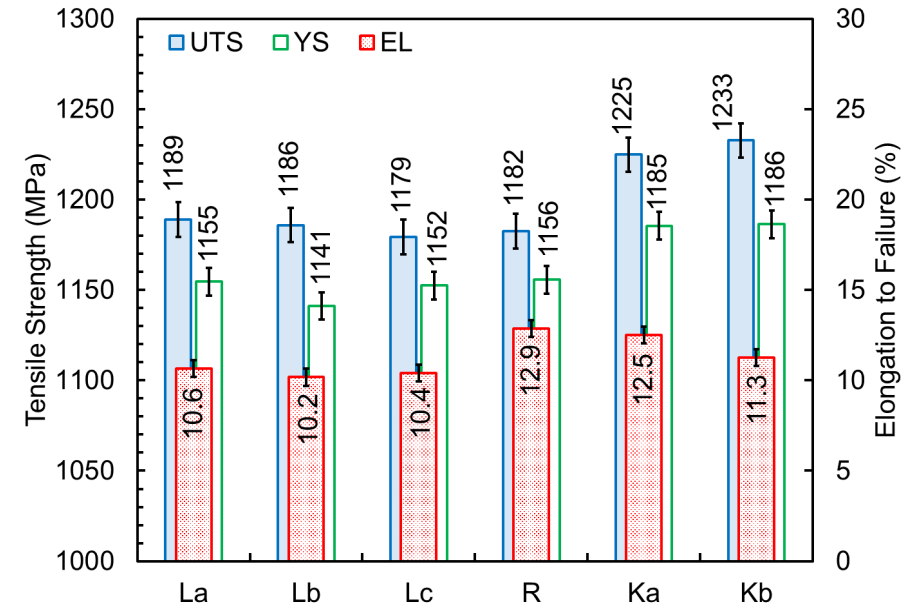
- 240 fatigue (16 x 15) and 96 tensile (16 x 6) specimens were fabricated
  - LoF: P-5%, P-10%, P-20%, H<sup>+5%</sup>, and H<sup>+20%</sup>
  - KH: P<sup>+30%</sup>V<sup>-20%</sup> and P<sup>+20%</sup>V<sup>-30%</sup>
- KH specimens were fabricated only in vertical orientation, while the recommended (R) and LoF ones were fabricated in vertical, diagonal, and horizontal orientations



# Tensile Properties

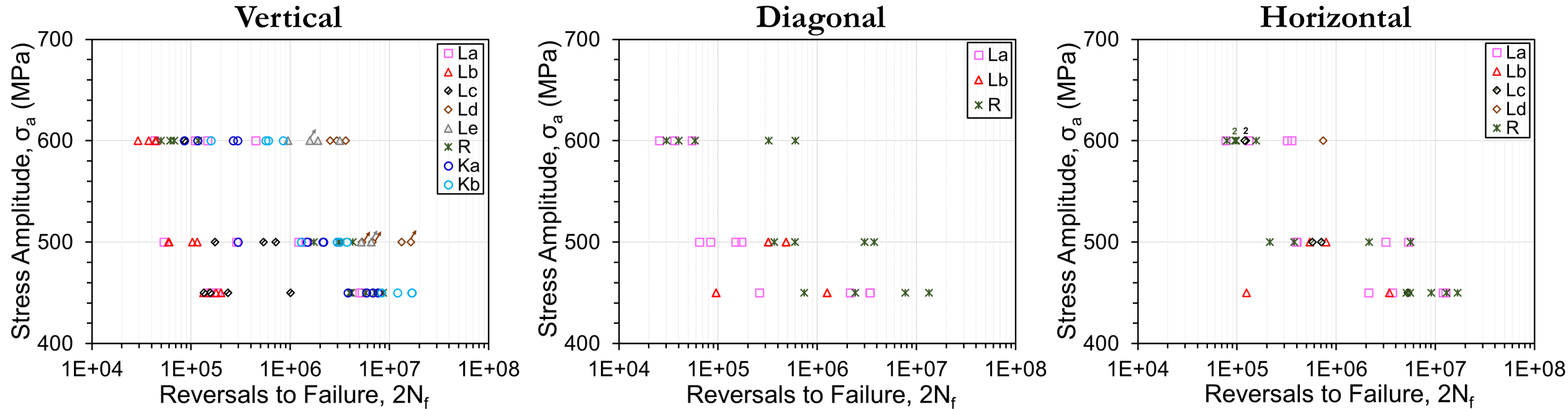


Note: 6 vertical tensile specimens were tested for each condition



- Yield strength (YS) and ultimate tensile strength (UTS) of all specimens were almost comparable
- KH specimens had slightly higher strengths which might be attributed to the higher nitrogen content due to excessive energy input during fabrication
- LoF specimens had lower ductility due to larger number and larger size of defects causing an early failure

# Fatigue Performance



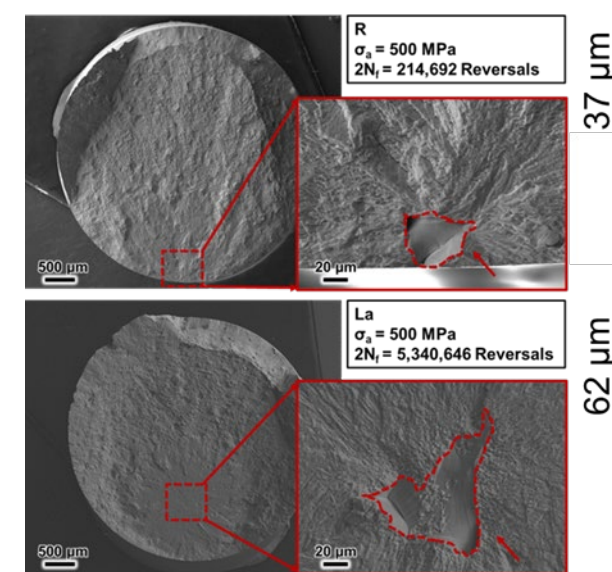
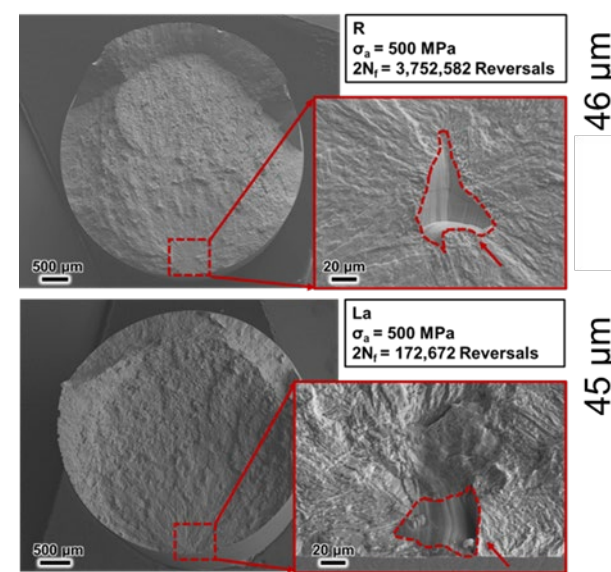
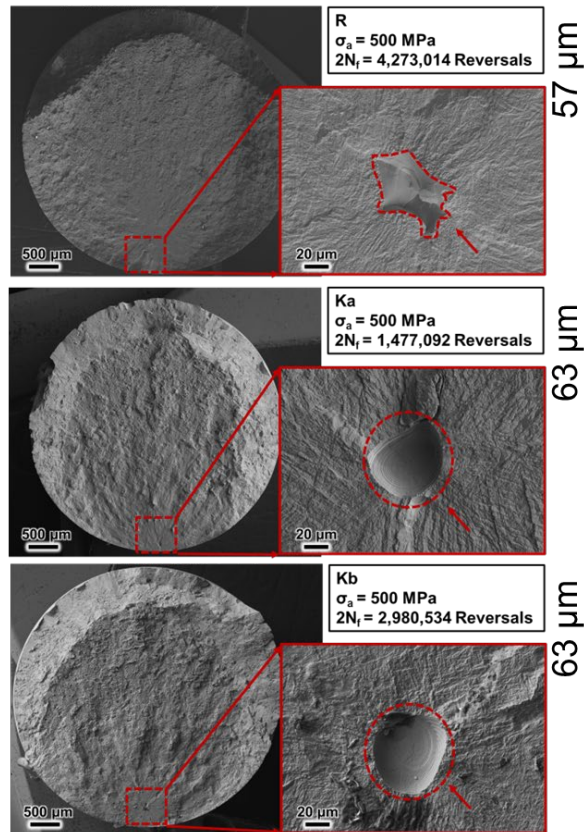
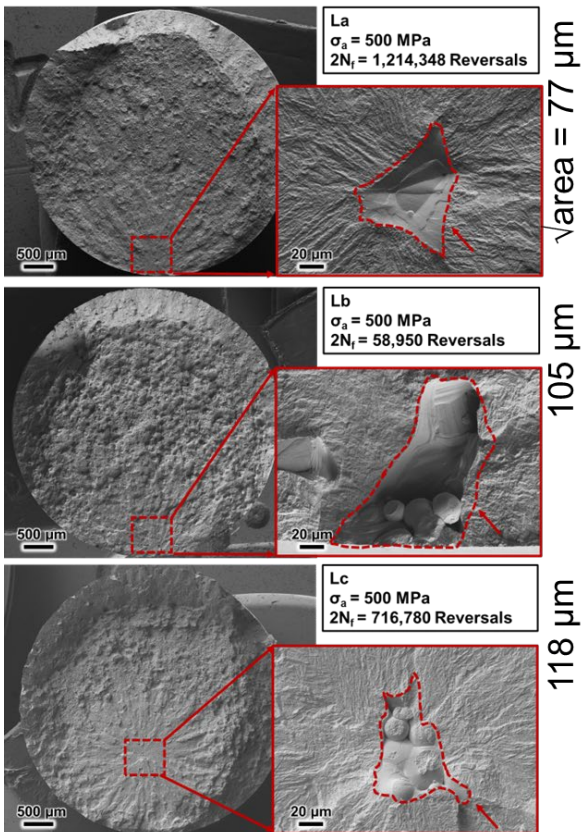
- In vertical orientation, KH specimens exhibited better fatigue performance than recommended ones
- LoF specimens exhibited worse fatigue performance for vertical and diagonal orientations
- Fatigue lives of LoF specimens had more scatter than KH ones due to wide variation in shape, size, and location of the crack initiating defects

# Fatigue Fractography

## Vertical

## Diagonal

## Horizontal



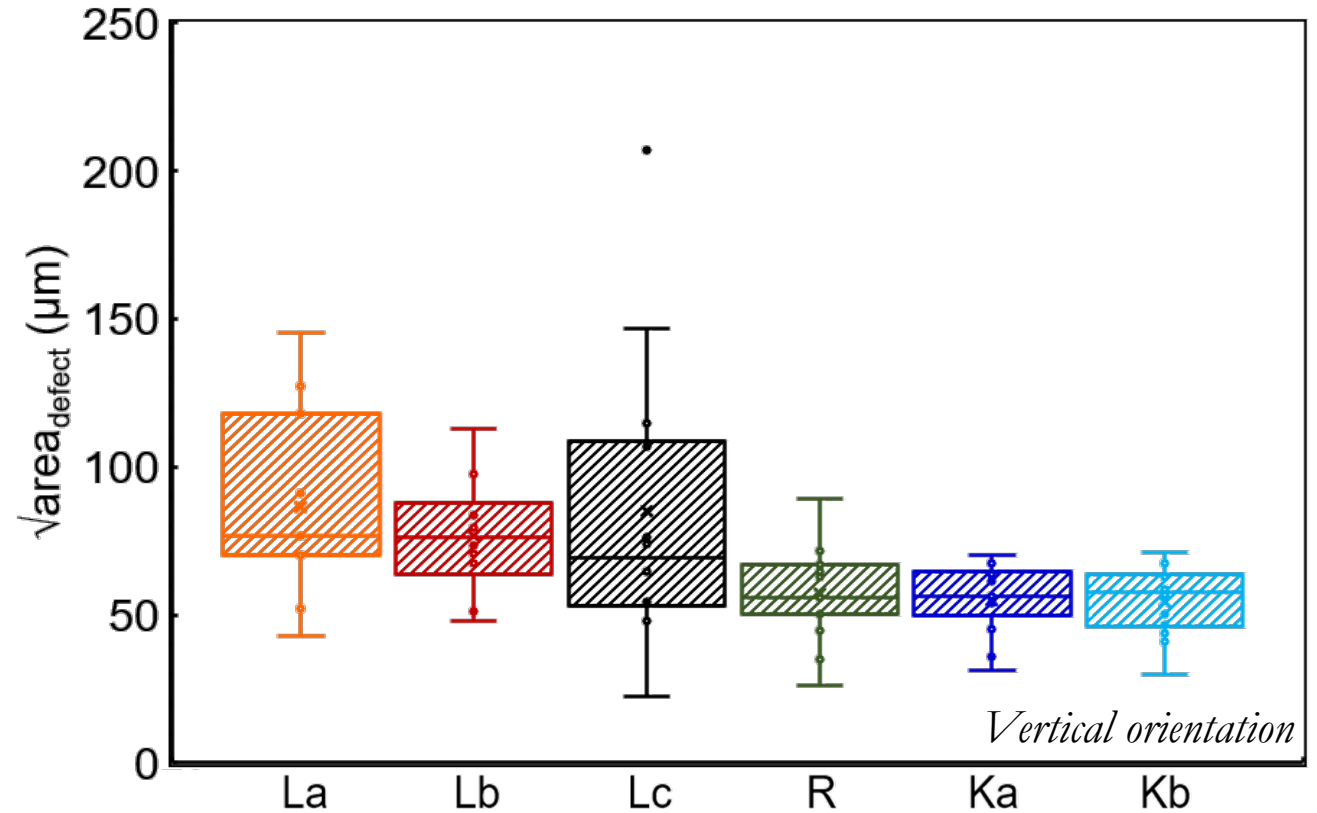
- LoF specimens: all fatigue cracks initiated from either internal or near surface LoF defects
- Recommended specimens: all fatigue cracks initiated from internal or near surface LoF defects
- KH specimens: fatigue cracks initiated mostly from KH defects and rarely from LoF defects located at internal or near surface

Note:  $\sqrt{\text{area}}$  of crack initiating defects is shown on the top right side of the fractography images



# Fatigue Behavior

- Defect sizes were measured using actual  $\sqrt{\text{area}}$  of the defect
- The size of the fatigue crack initiating defects of recommended and KH specimens were comparable
- Mean of the crack initiating defects of LoF specimens were significantly larger compared to recommended and KH specimens
- Size of the defects explained the order of fatigue life



# Summary

- Defect types shared different degrees of overlaps in the ranges of their morphological parameters
- The defect classification methodology could provide >98% and >99% accuracy when implemented into decision tree and artificial neural network, respectively
- KH specimens exhibited slightly higher tensile strength due to higher amount of nitrogen
- KH specimens exhibited better fatigue performance due to smaller crack initiating defect sizes and slightly higher tensile strengths
- LoF specimens exhibited scatter in fatigue behavior due to differences in crack initiating defect sizes

# Thank you for your attention !

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