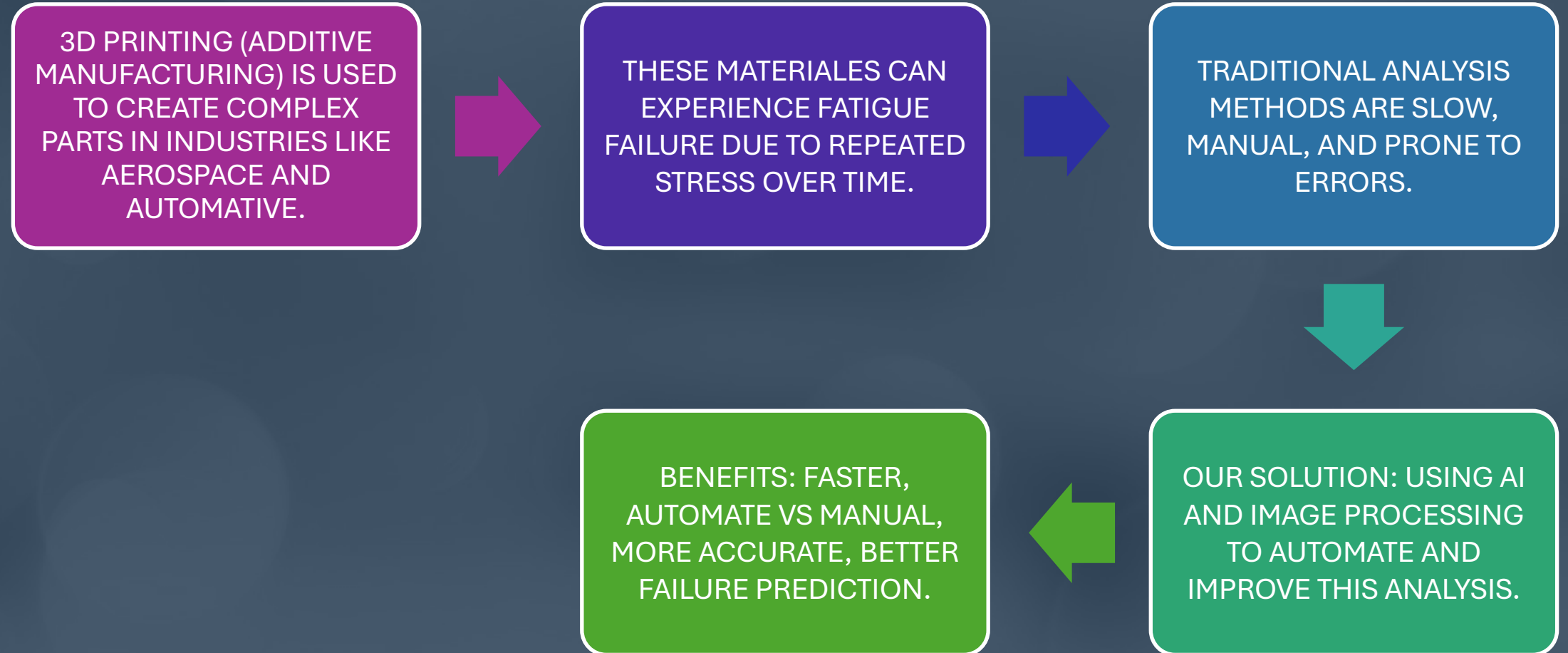


Using AI and Image Processing to Improve Fatigue Failure Analysis in 3D-Printed Materials

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Avichay Mazin

Introduction

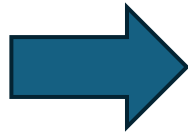
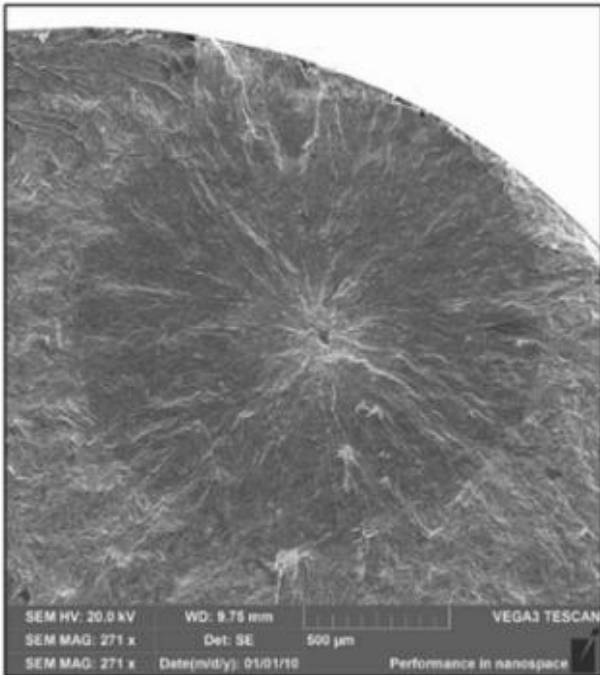


Additive Manufacturing & Fatigue Failure

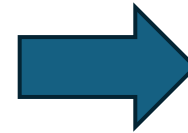
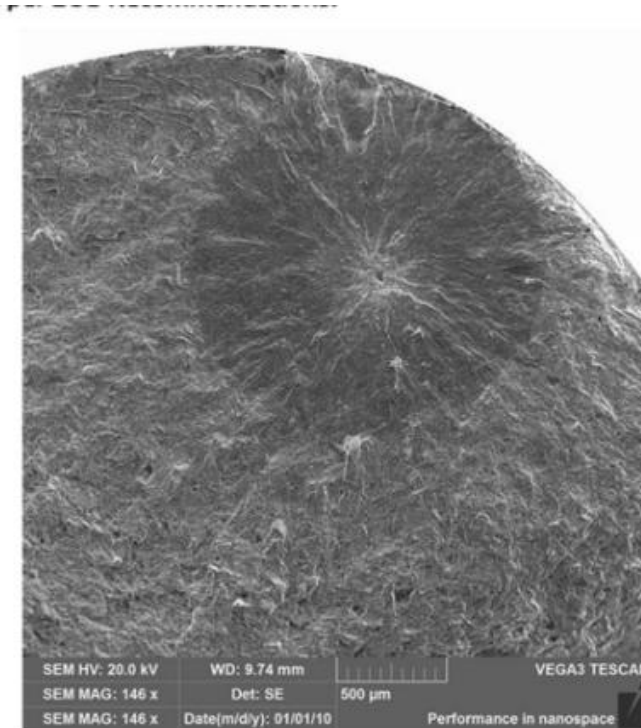
- **What is Additive Manufacturing (AM) ?**
 - 3D printing technique used to create parts layer by layer.
 - Popular in aerospace and automotive for lightweight and complex designs.
- **Fatigue Failure in Materials:**
 - Repeated stress leads to material failure (cracks, fractures).
- **Challenges with Traditional Analysis:**
 - Involves **manual fractography**, where experts visually examine fracture surfaces.
 - Time-consuming, prone to human error, and lacks scalability.
 - Difficult to predict when and where cracks will occur.

Failure Propagation Phases

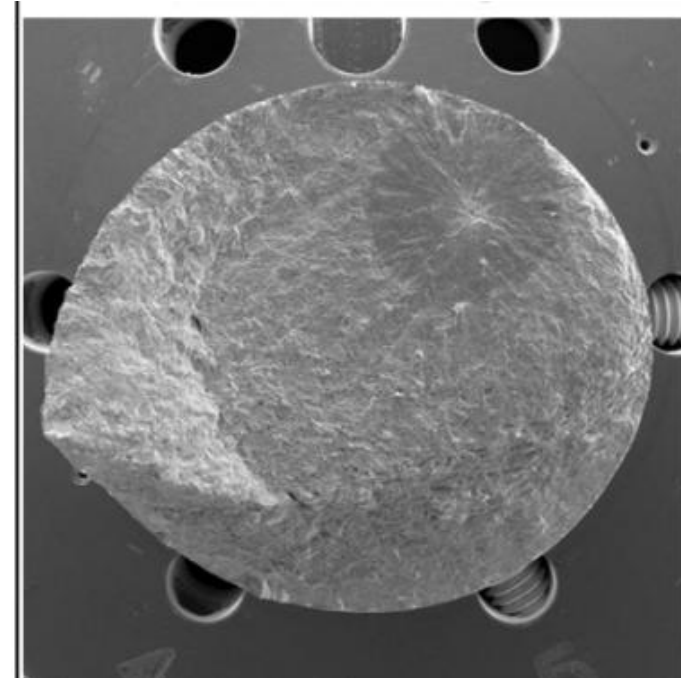
Initiation Source



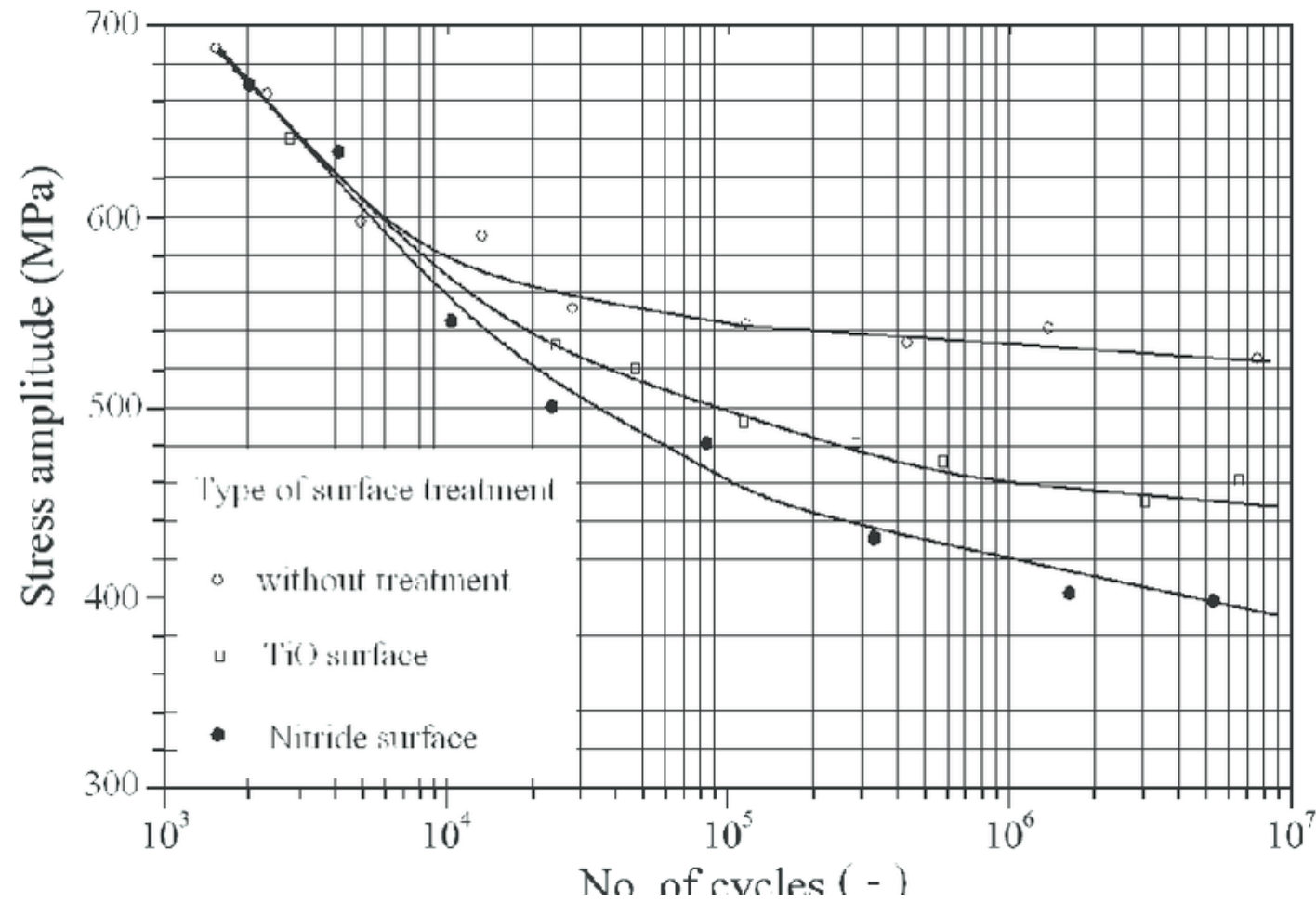
Crack propagation



Final Failure



Fatigue Failure



Dataset

- The dataset utilized in this research comprises 63 Scanning Electron Microscope (SEM) images of Ti-6Al-4V alloy fatigue specimens.
- The dataset is organized into three categories based on the print quality recommended by the manufacturer: P1, P2, P3 ...
- Included of duplicated images for certain specimens from lower and upper surfaces



Figure 3: Fatigue (crack initiation) & Crack Growth Test Specimens and the Tensile Machines used

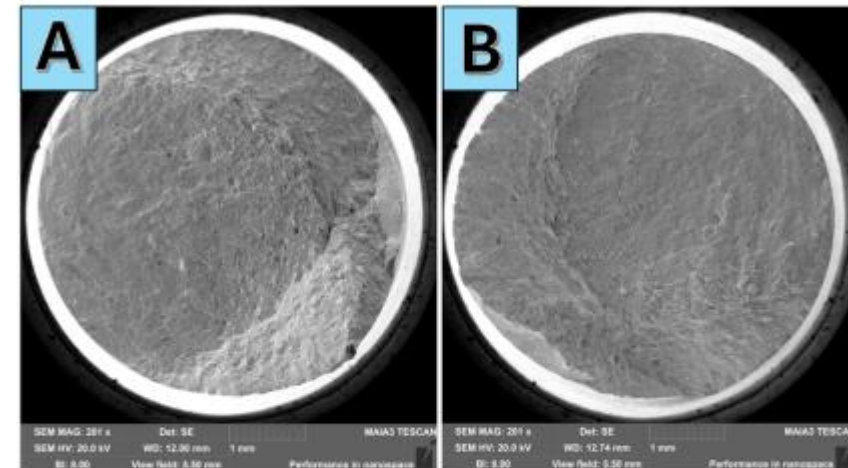


FIGURE 3: SEM images of a Ti-6Al-4V fatigue specimen from two opposing viewpoints (A and B), showcasing the fracture surfaces from the lower and upper surfaces at different angles.

AI and Image Processing Approach

Our solution:

- Combine image processing techniques and AI to analyze fatigue failure in 3D printed materials.

Key steps:

Image Proccesing:

- Use sobel operator to detect edges for generate heatmaps for identifying critical features (crack ares).

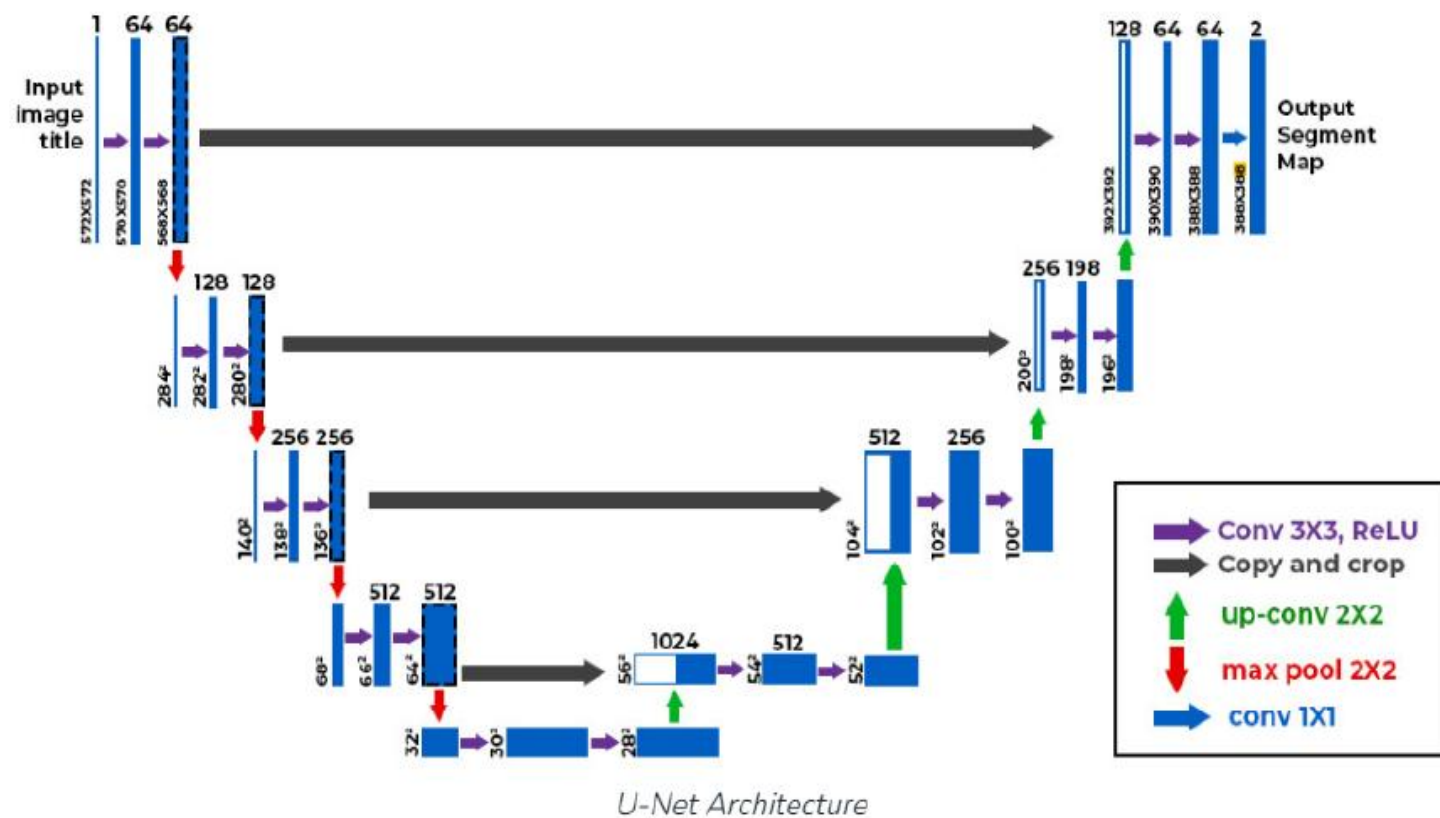
Preproccesing:

- resize and clean Scanning Electron Microscope (SEM) images for uniform analysis.

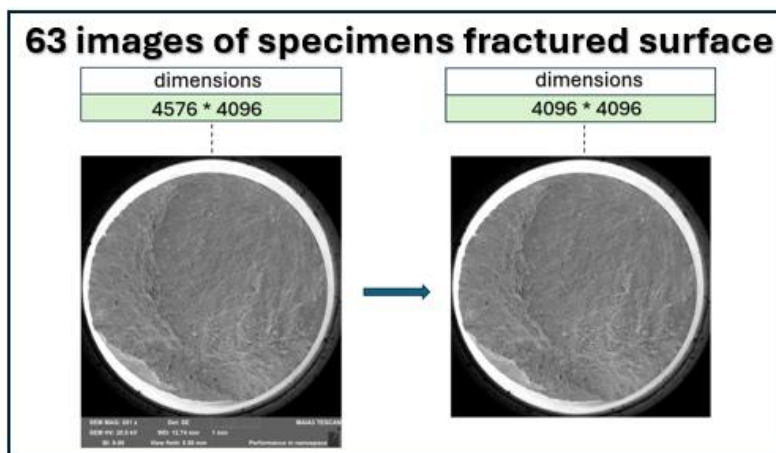
AI (Unet model):

- Apply **UNet architecture** for image segmentation.
- Detect external contours (whole surface) and internal contours (crack propagation) in fracture surfaces.

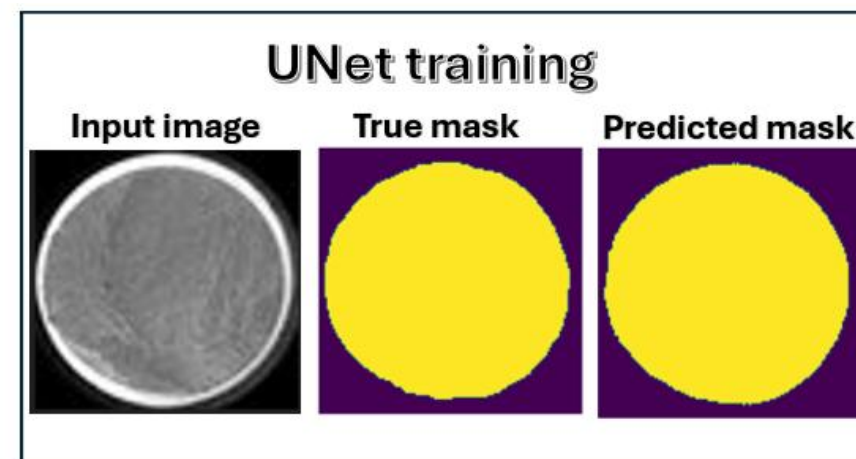
Proposed approach - U-Net



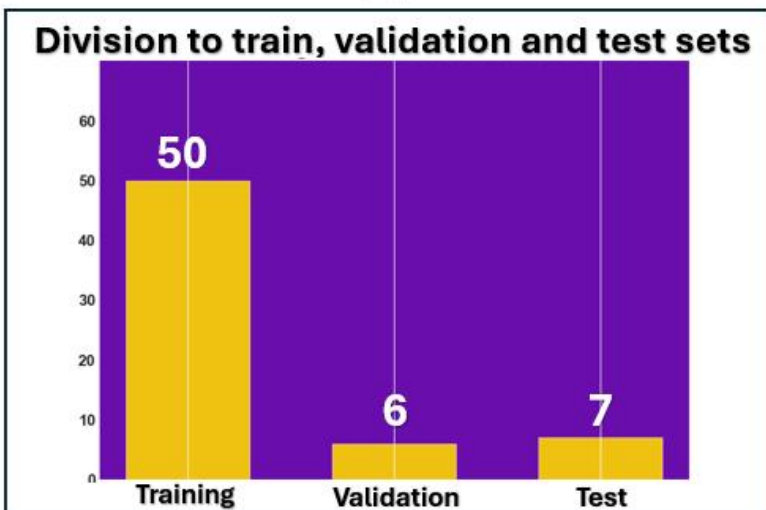
Visualization of the Process



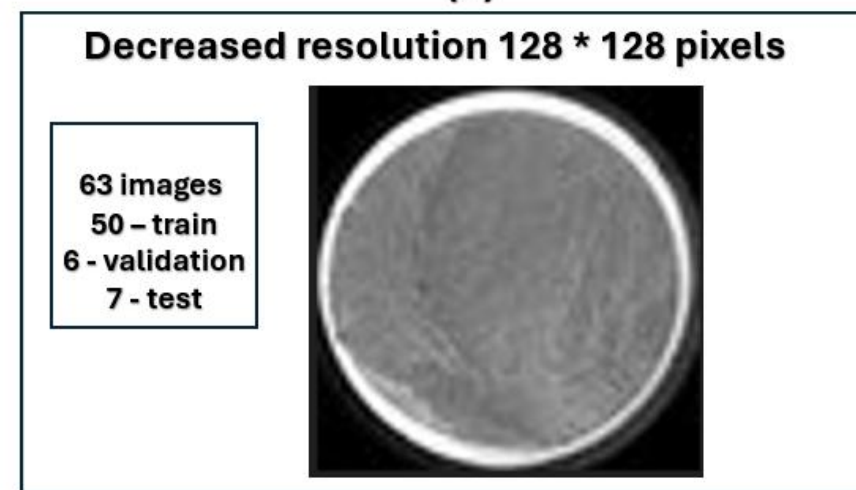
(a)



(d)

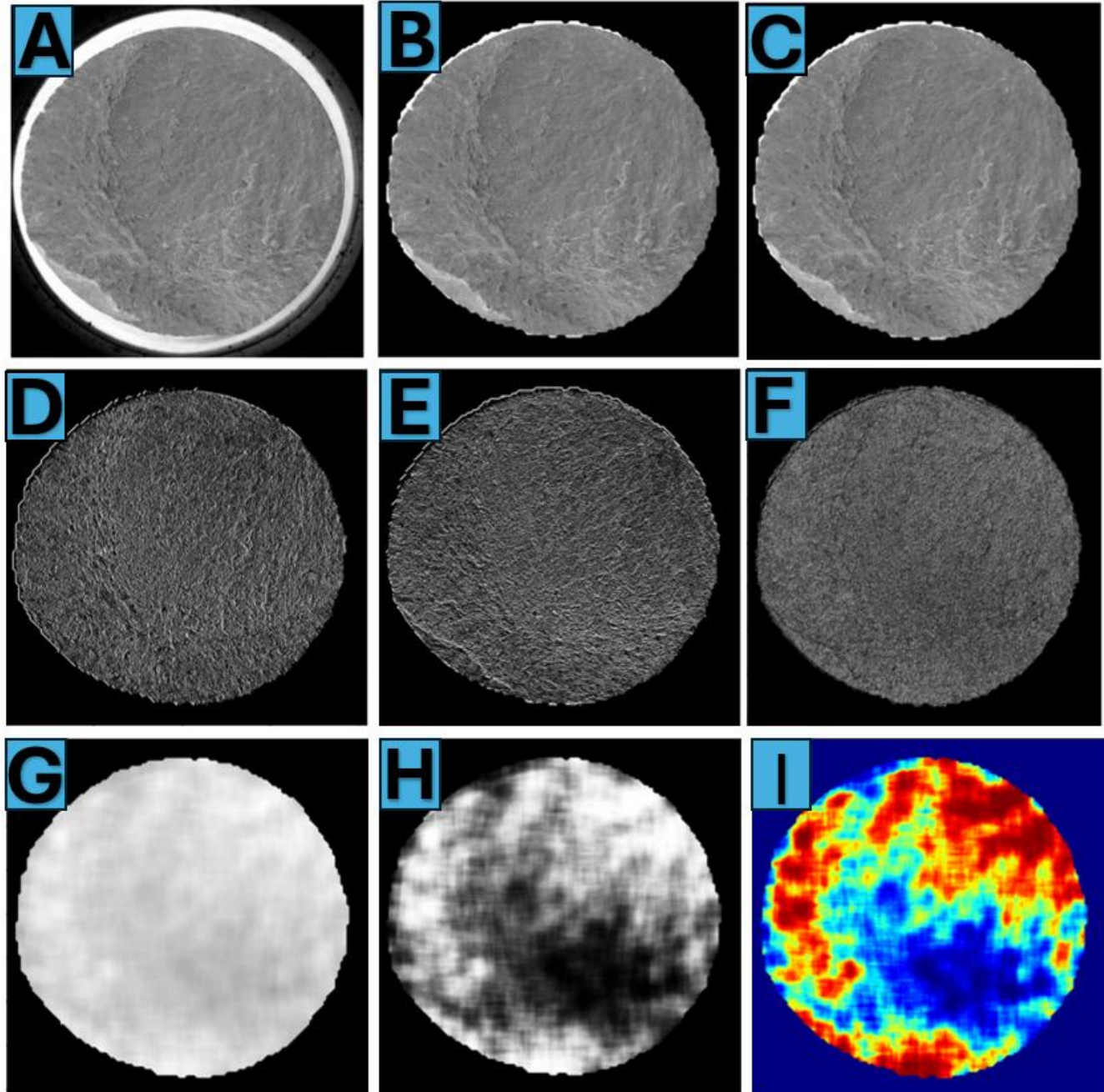


(b)

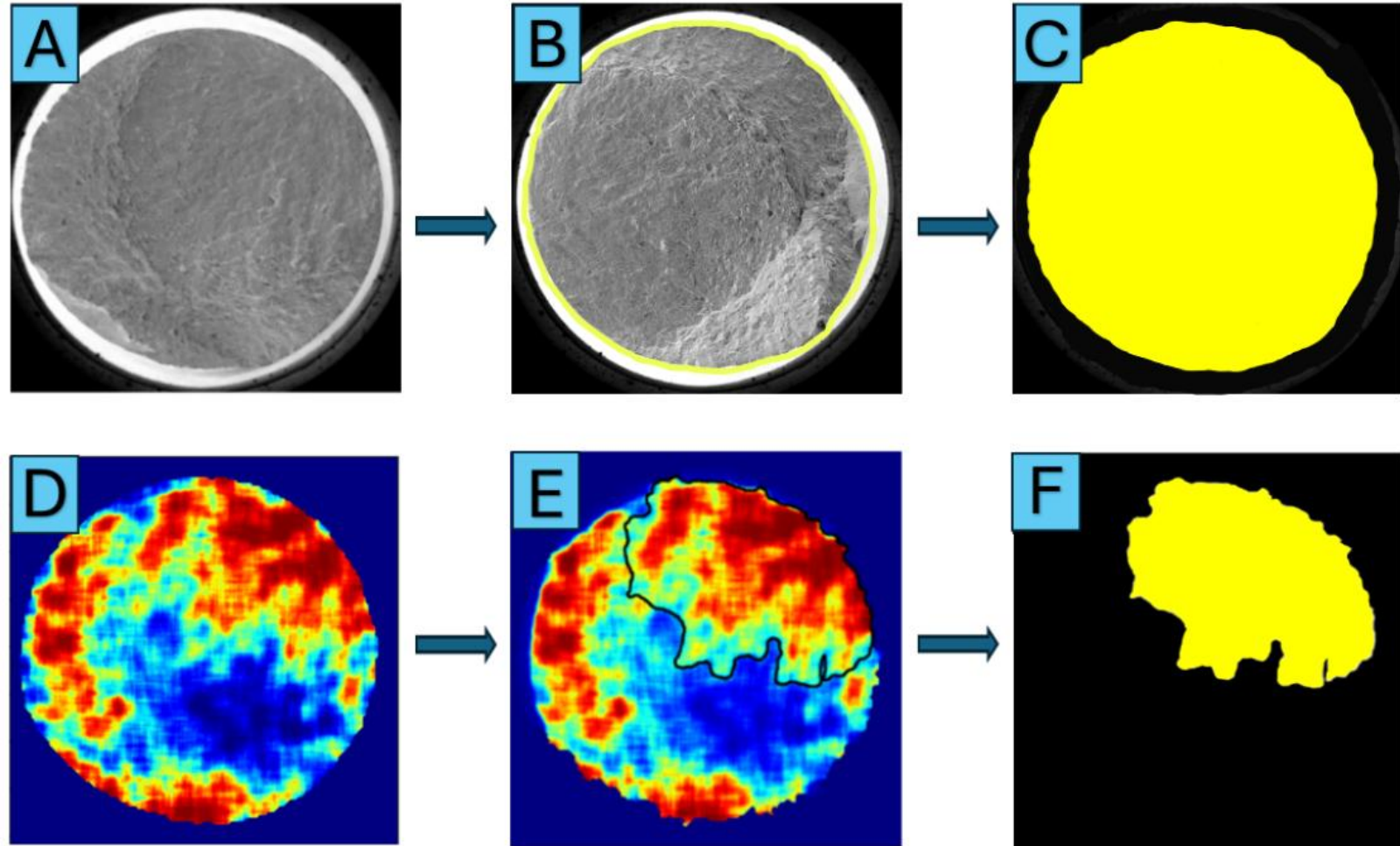


(c)

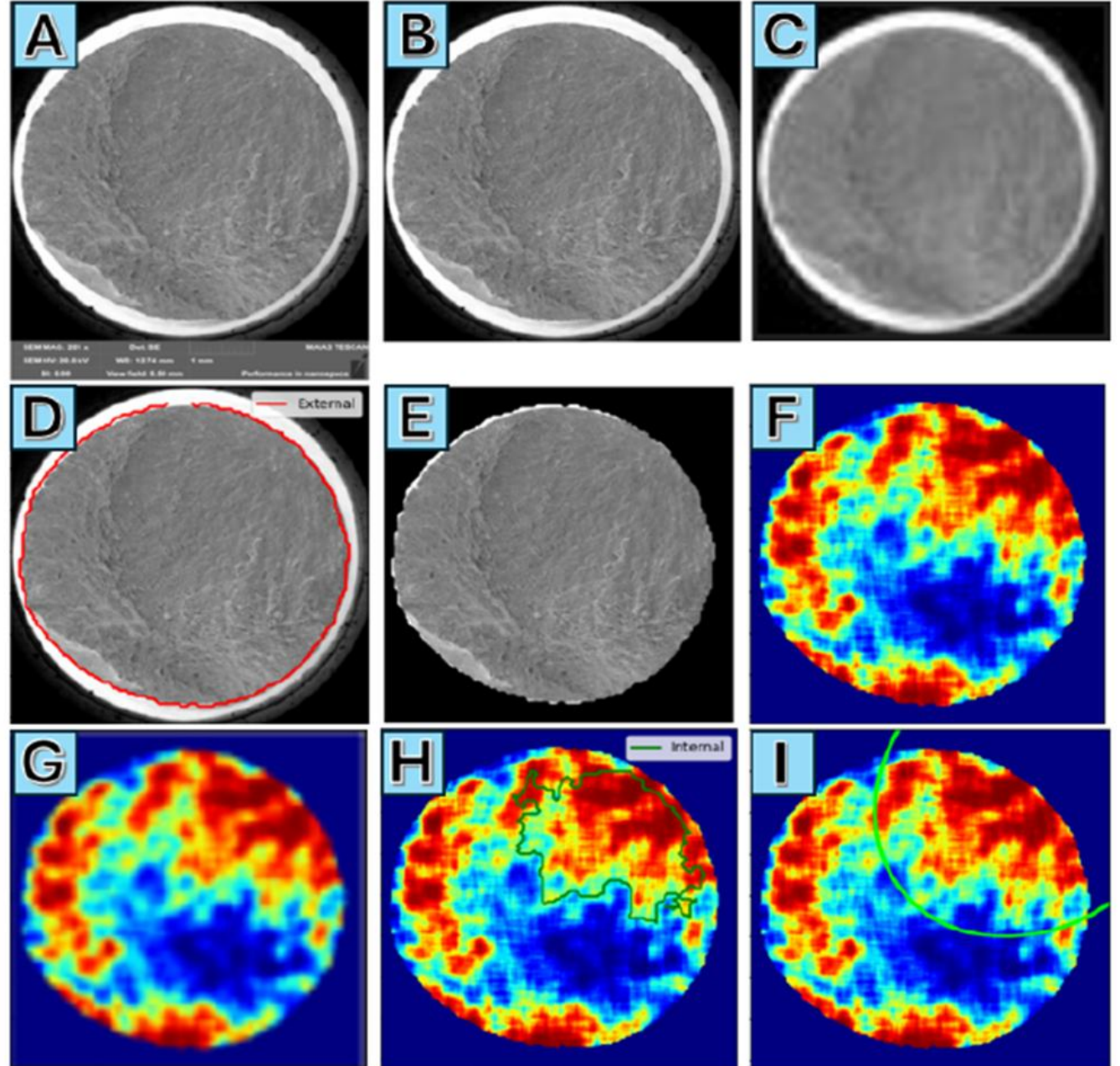
Visualization of the Process- Heatmap using sobel operator



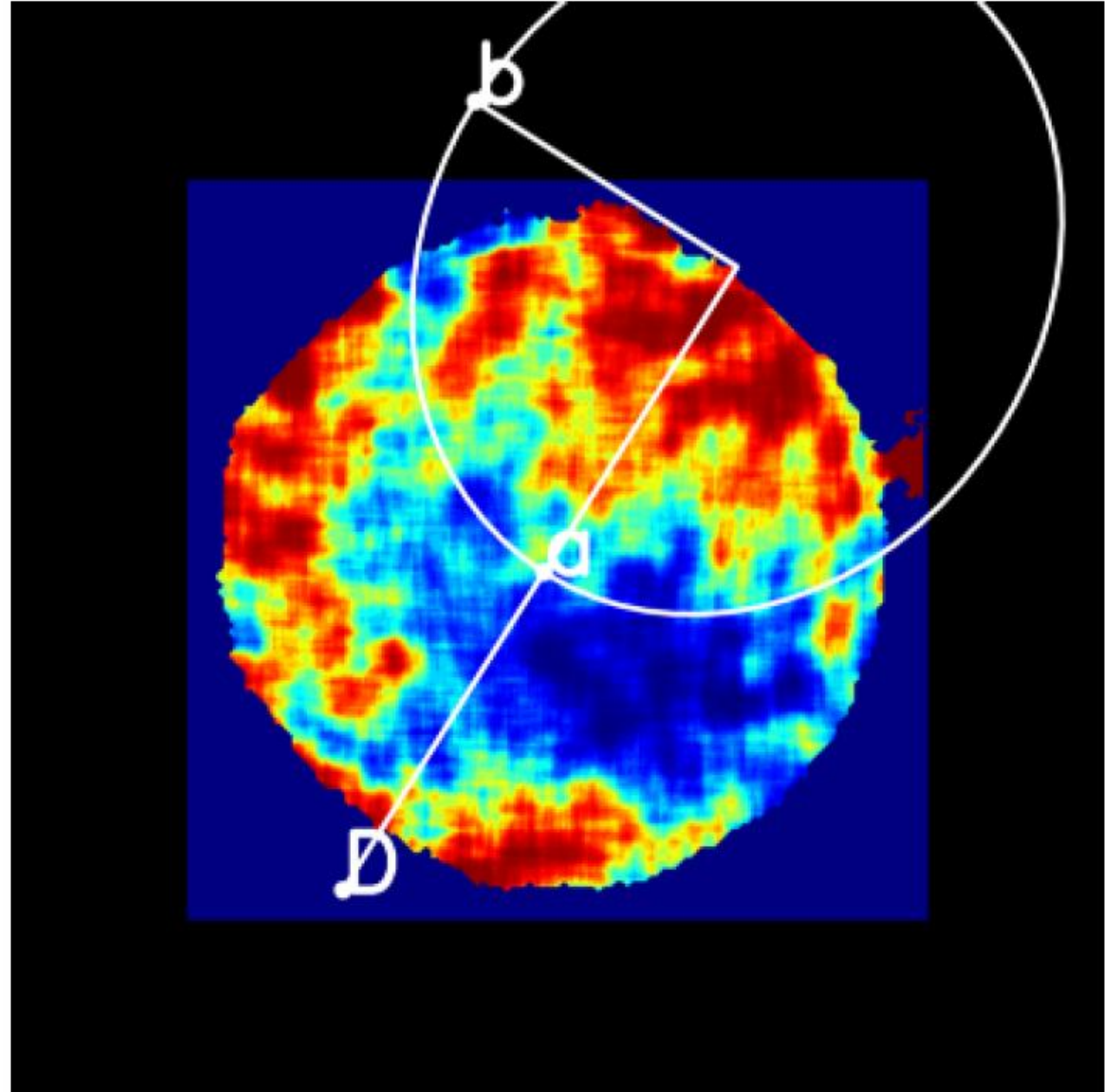
Visulaization of the Process – Labeling



Visualization of the Process – Get ellipse



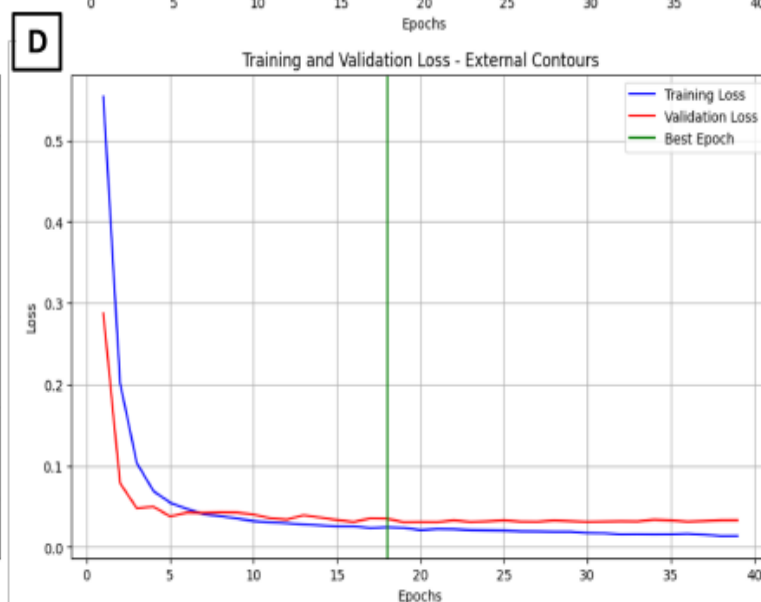
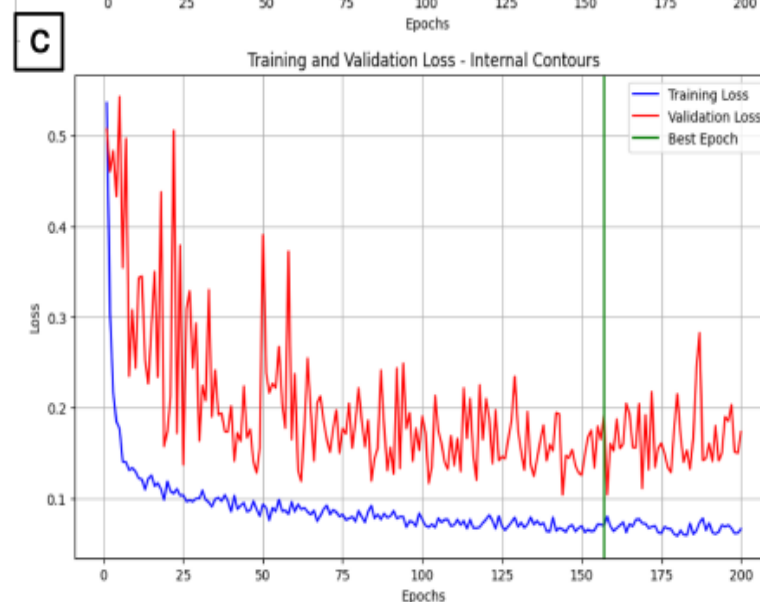
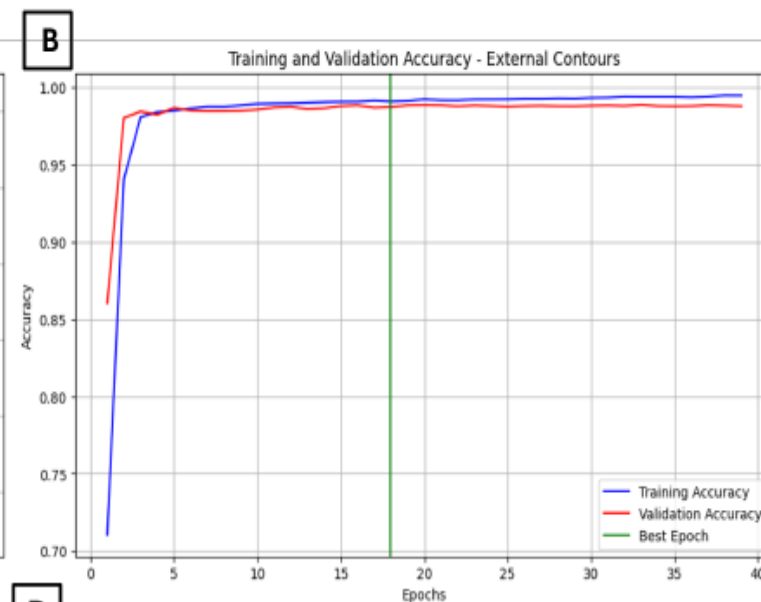
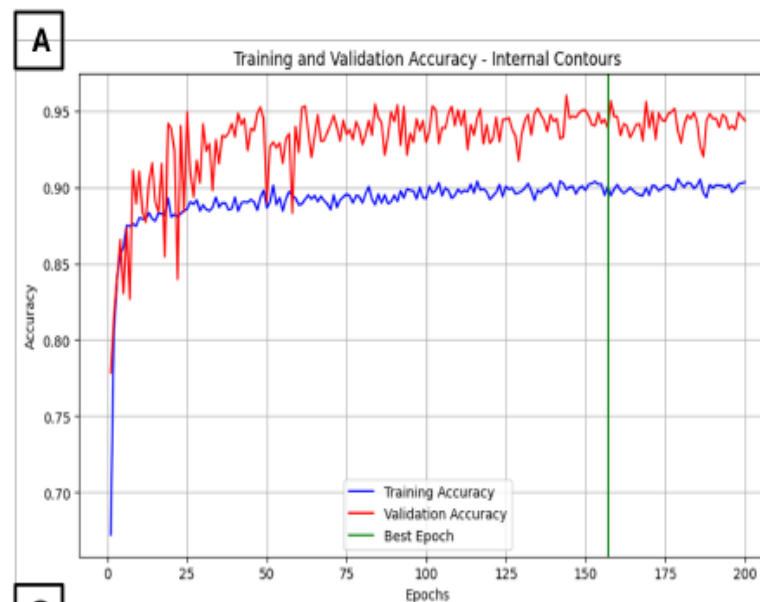
Visualization of the Process – Get ellipse



Results

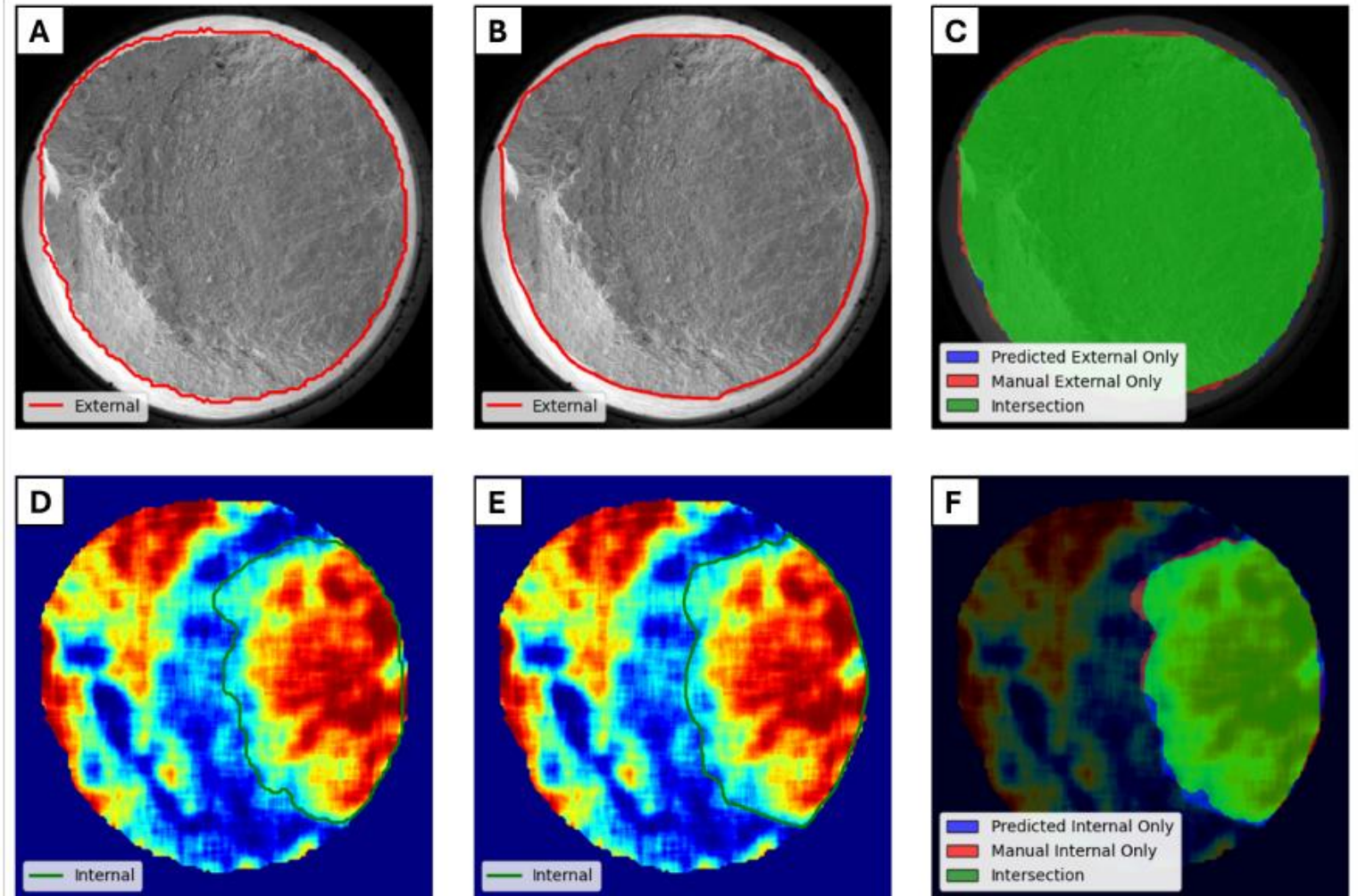
Training and
Validation
accuracy and loss

—
External and
internal contours



Results

IOU -
(Intersection
Over Union)



External contour detection achieved 0.97 in the IOU.
Internal contour detection achieved 0.93 in the IOU.

Conclusion and Future Work

Conclusion:

- Developed an AI-driven solution for fatigue failure analysis in 3D printed materials.
- Achieved high accuracy: IoU of 0.97 for external contour and 0.93 for internal.
- Faster and more reliable than traditional manual methods.

Future Work:

- Extend the model to analyze other materials and failure types.
- Explore real-time monitoring during 3D printing processes.
- Investigate additional AI models for even better accuracy and speed.