



GO BEYOND

POTENTIAL INDUSTRY STANDARD APPROACH TO MEASURE AND QUANTIFY MICROSTRUCTURES AND MICROTEXTURE IN TITANIUM

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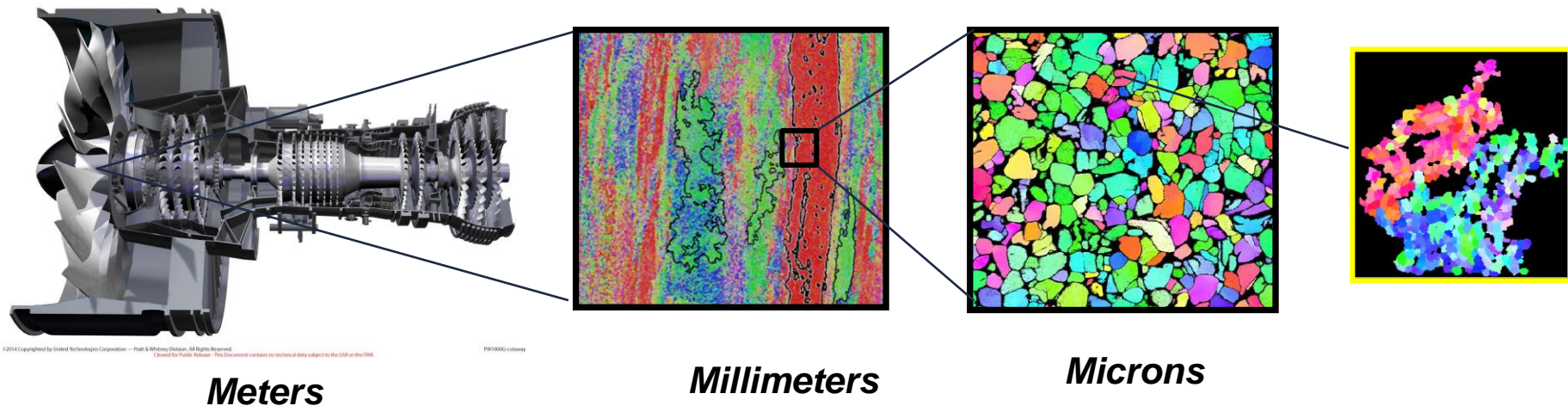
AGENDA

LET'S GET STARTED

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Cold Dwell Fatigue and Microtexture	04
Processing Effects on Microstructure	05
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BACKGROUND

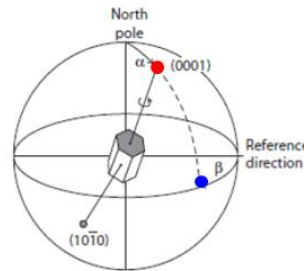
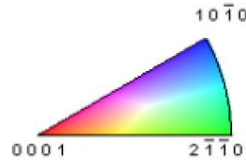
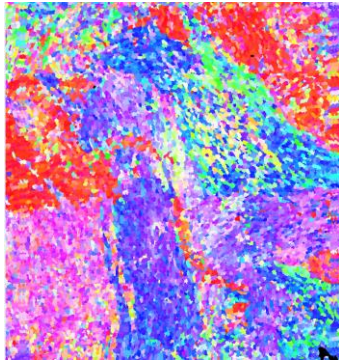
- Titanium is a work-horse material in aerospace industry
- As applications for titanium increase, enhancements and control of microstructure must evolve
 - Case study: High cycle fatigue improvement through microstructure control
 - In progress: Low cycle (dwell) fatigue improvement through microstructure + texture control



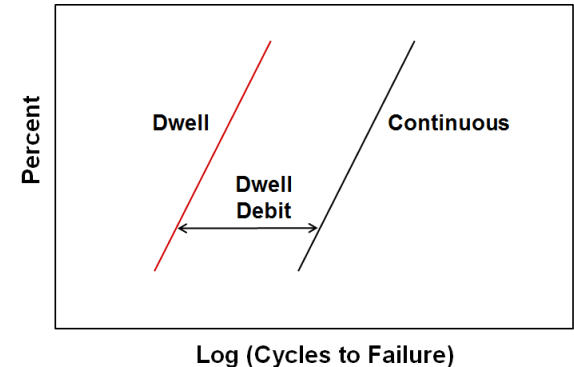
Need industry standard methods to quantify microstructure and microtexture in titanium

MICROTEXTURE AND DWELL FATIGUE

- **OIM (Orientation Image Microscopy)** provides crystal orientation information
 - Not possible to do quantitative assessment by eye
 - Not all textured regions impact component fatigue properties equally
- **“Texturing” or “Micro-Textured Region (MTR)”**
 - Small, localized regions where grains are aligned in similar direction
 - High stresses can develop at grain boundaries of hard regions leading to crack nucleation
 - High levels of microtexture can lead to significant dwell debit



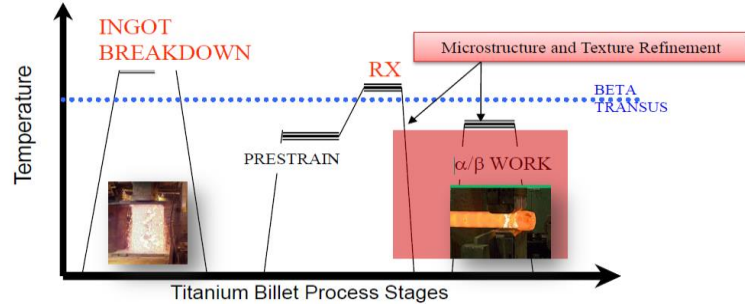
Example of Normal Projections for HCP Crystal



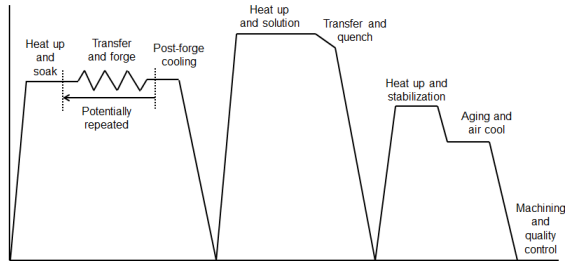
Dwell fatigue capability heavily influenced by microtexture.
Need standardized methods for its control in titanium product.

MANUFACTURING EFFECTS ON MICROSTRUCTURE

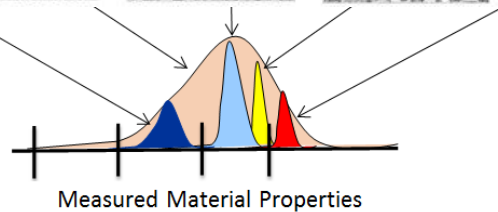
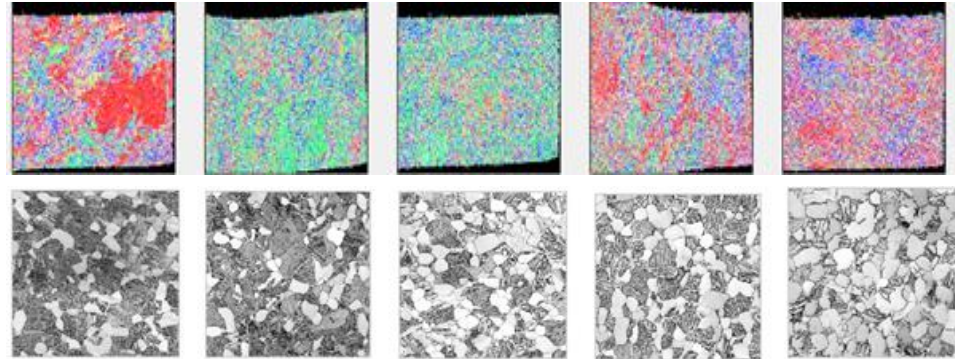
Billet Conversion Process



Forging + Heat Treat



Microstructure Variability

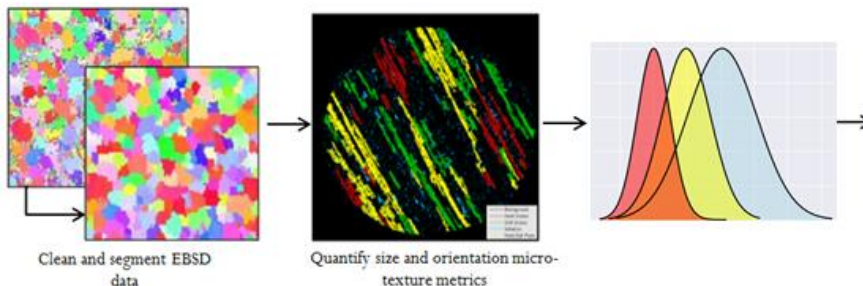


Microtexture and microstructure is process-dependent – not random.
Standardized methods have been developed quantify and model evolution during processing.

POTENTIAL METHOD FOR MICROTEXTURE QUANTIFICATION



DREAM.3D

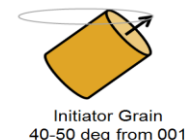
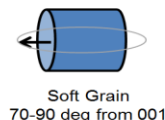
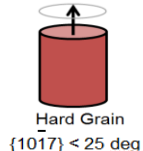
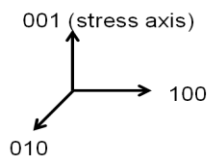


- ☐ A / B comparisons
- ☐ Quality Control
- ☐ Property Prediction
- ☐ Life Assessment

Key Input Parameters:

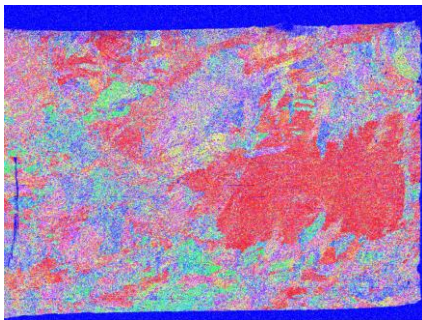
- Pixel C-Axis Misalignment Tolerance
- Minimum Size for MTR

Geometric: <ul style="list-style-type: none">• MTR Size (μm^2)• Aspect ratio• Solidity (%)• MTR Intensity• MTR Density (#/area)	Orientation: <ul style="list-style-type: none">• Avg. Misorientation (<u>deg</u>)• Stress axis misalignment (<u>deg</u>)• MTR class label	Neighbor Interactions: <ul style="list-style-type: none">• Spatial relations/frequency• Hard/Soft Pair Intensity• Pair Contact Lengths (μm)
Total Metrics: 76 per scan		

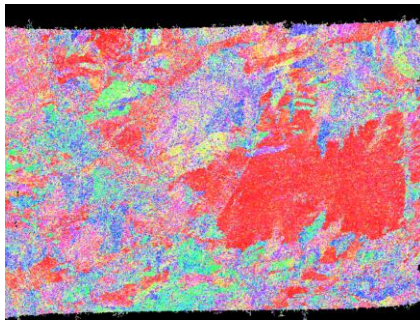


DREAM3D is free, open-source software that is well-suited for OIM data analysis.
Prevalent use in aerospace industry for quantification of microtexture.

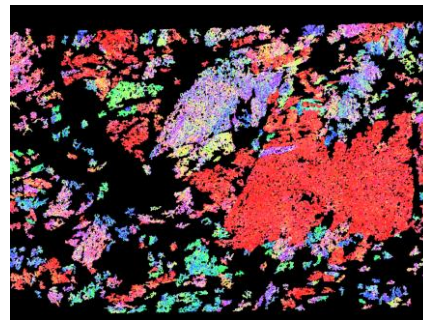
POTENTIAL METHOD FOR MICROTEXTURE QUANTIFICATION



RAW DATA

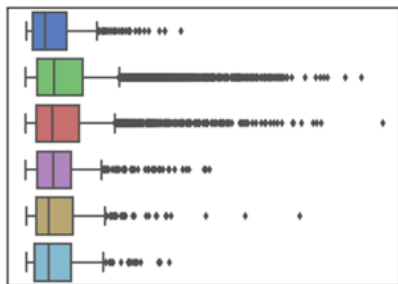


CLEANED DATA



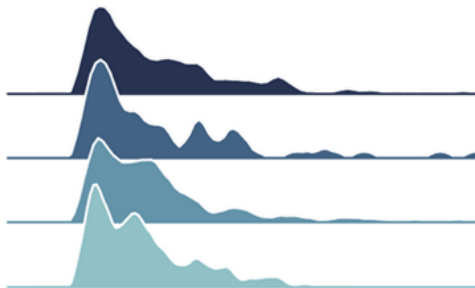
MTRS ONLY

Compare MTR Metrics for
Different Pedigrees of Material



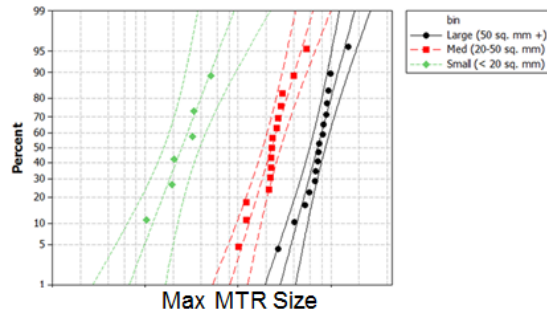
MTR Size

MTR Size Histograms



MTR Size

Probability of Finding Large MTRs
as a Function of EBSD Scan Size

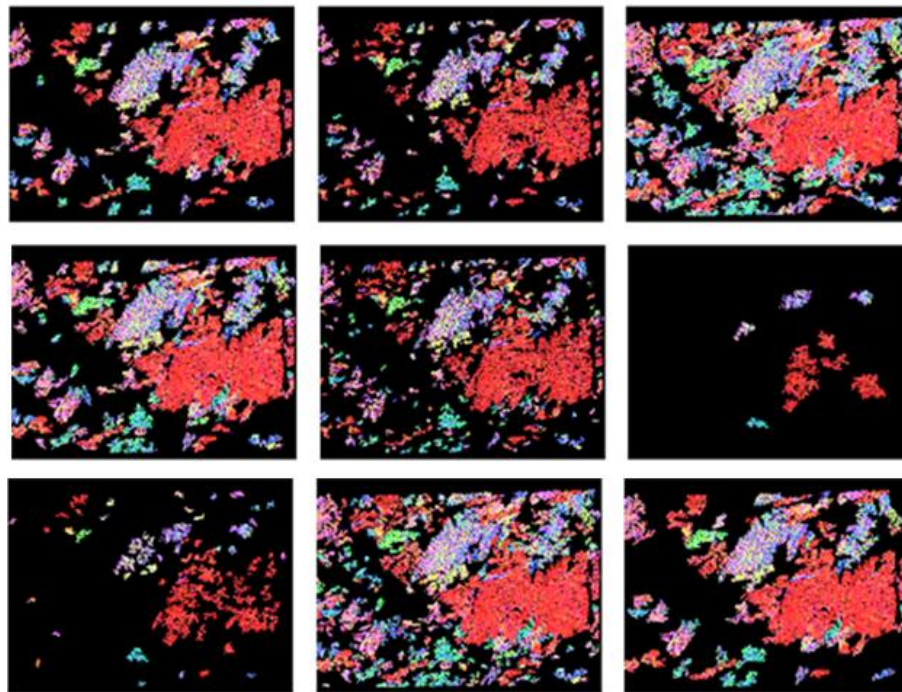
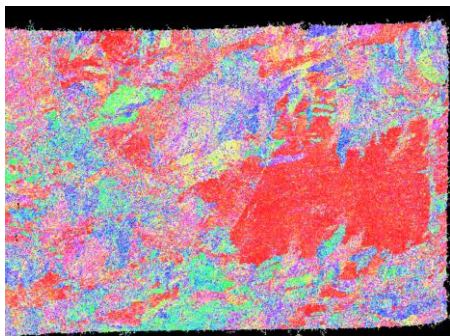


Dream3D enables statistical analysis and comparison of microtexture data

POTENTIAL METHOD FOR MICROTEXTURE QUANTIFICATION

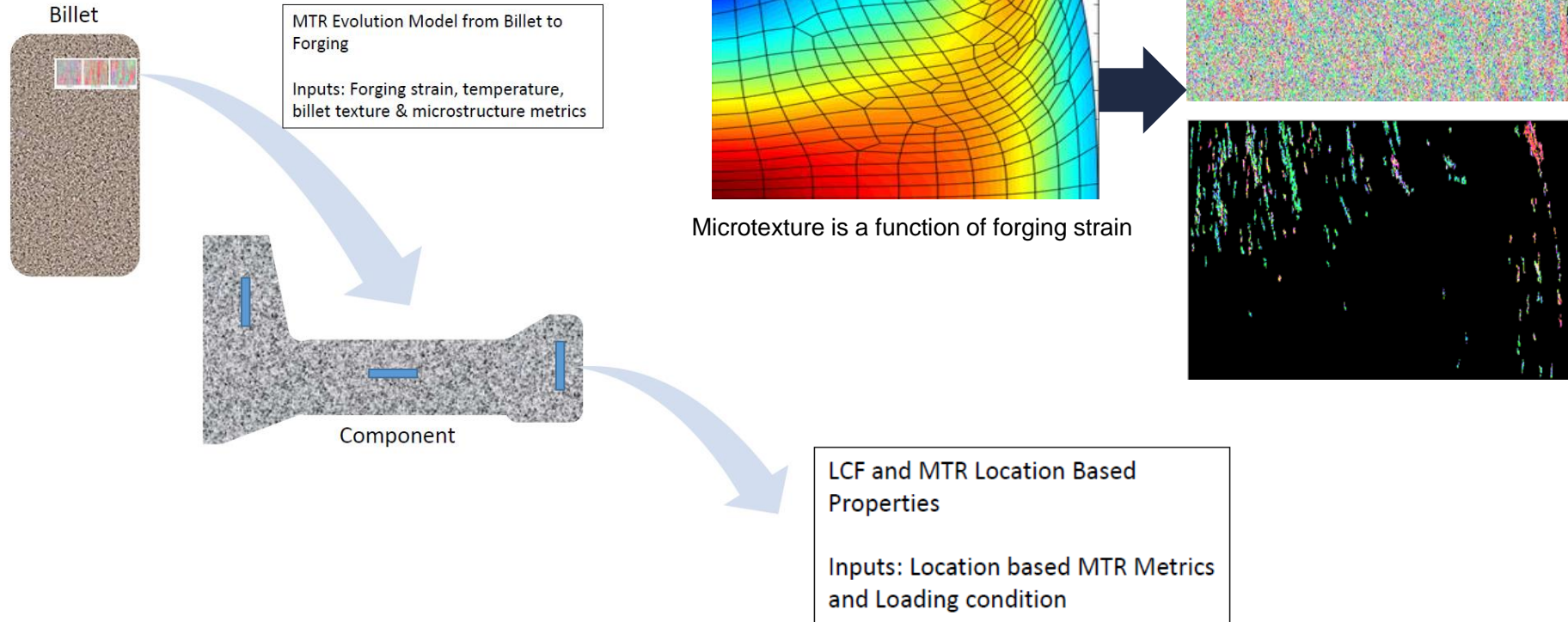
OUTPUTS USING DIFFERENT DREAM3D SEGMENTATION PARAMETERS

CLEANED EBSD DATA



Analysis of OIM data is sensitive to certain Dream3D parameters and can significantly affect results. Industry standards improve ability to compare results and develop quality criteria.

FUTURE STATE



ICME approach to predicting microtexture evolution during manufacturing and location-specific properties will help define billet and forging level criteria to minimize effects of microtexture

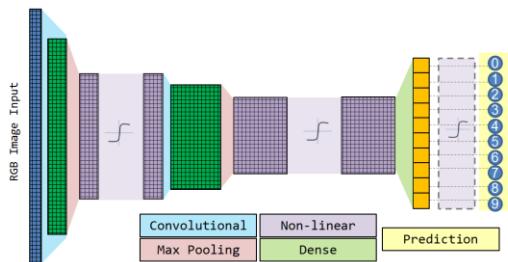
LONG TERM CHALLENGES IN MATERIALS CHARACTERIZATION

Current state:

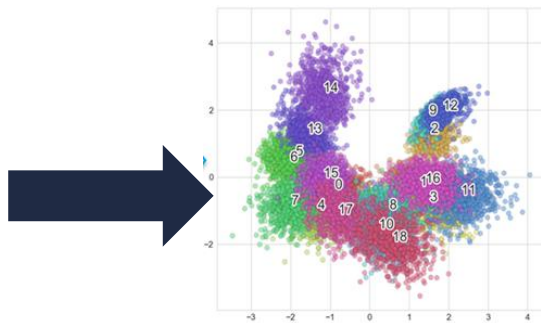
- Volume of data collected is increasing rapidly (automated scopes, new methods, etc)
- Current models based on simple metrics like grain size or volume fractions

Viable options moving forward:

- Fast, scalable, accurate, flexible techniques are needed
- Examples: machine-learning or image-based neural networks



Convolutional neural networks (CNN)
(enable image recognition)



Predicting material pedigree or
properties using CNNs

Applications & Benefits:

- ✓ Fast (analyze 1000's images per second)
- ✓ Repeatable, transferrable
- ✓ Automatically learn important features
- ✓ Predict material pedigree from images
- ✓ Image visual similarity lookup
- ✓ Quality control (defect recognition)

Machine learning and AI are enabling rapid materials characterization and improved modeling capability

CONCLUSION

- A method to quantitatively characterize texture has been developed.
- The developed method has potential to be an INDUSTRY STANDARD
- In order for titanium to be applied to future demanding applications, further control of microstructure and microtexture will be required

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 - Air Force Research Lab
 - Arconic
 - Wyman Gordon
 - Materials Resources, LLC



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