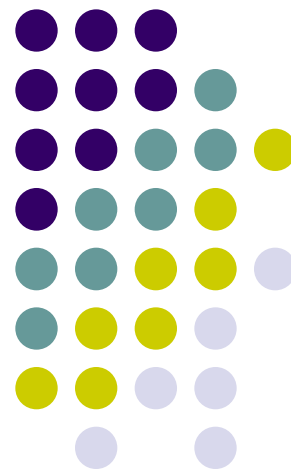




Introduction to Data Workflows in ICMSE

May 12, 2015

Vasisht Venkatesh
David Furrer



2015 UCSB Workshop on Collection and Analysis of Big Data in 3D Materials Science

University of California – Santa Barbara
May 12th-14th 2015

No Technical Data per the EAR or ITAR

Data Workflows in ICMSE



- Materials and Process Development
- Material development
 - Data management plans
 - Characterizing controlling microstructural features
 - Data schema
- Model Development and Validation
 - Verification and Validation
 - Bayesian updating
- Production Data
 - Material and Model Monitoring
 - Model Enhancement



Additive Manufacturing: rapid, low cost, flexible



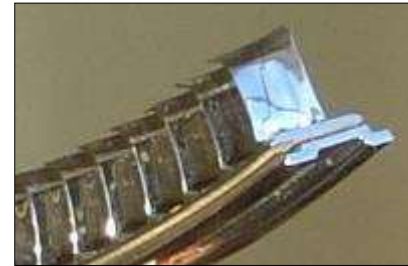
Freeform Fabrication of Parts

- Shorter Lead Times
- Eliminate Tooling

Alternatives to Castings

New Innovative Designs

Powder Bed Technologies



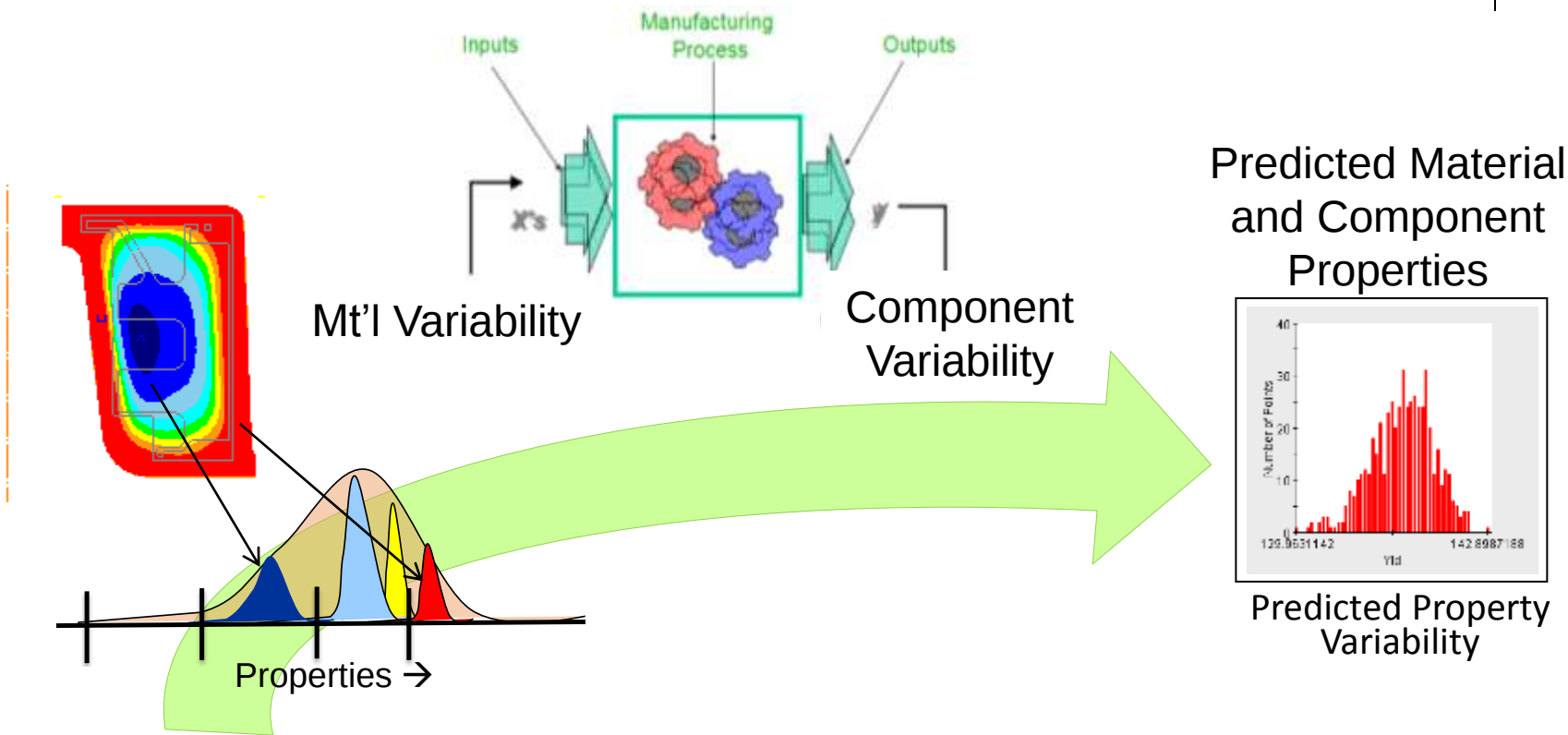
Direct Metal
Laser Sintering (DMLS)



Electron Beam
Melting (EBM)



Predicting Material Capabilities



Measured Material Properties

Controlling Feature and Mechanism MUST be Understood

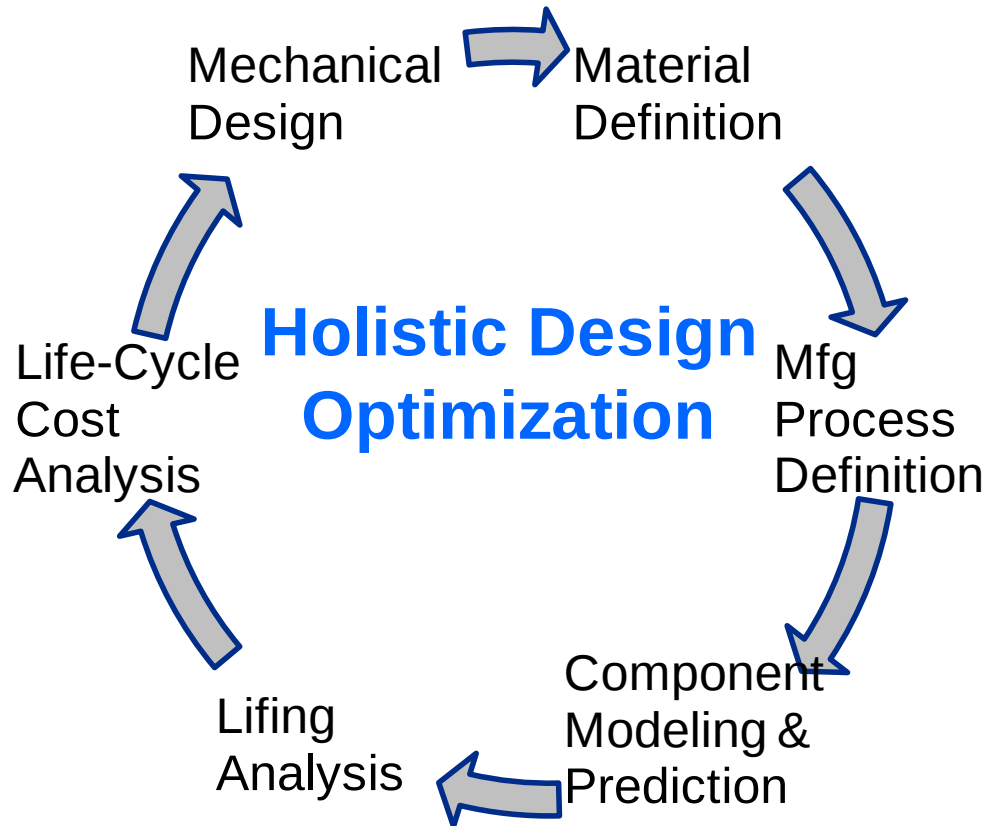


No Technical Data per the EAR or ITAR

2015 UCSB Workshop on Collection and Analysis of Big Data in 3D Materials Science

Computational Materials Modeling

Fit for purpose focus



Materials Modeling:

- Reduced cycle-time
- Capability optimization
- Enhanced material definition

Mfg Process Modeling:

- Reduced cycle-time
- Increased yield and product quality

Component Modeling:

- Location-specific optimization
- Optimize part weight/cost
- Maximized lifing capability

Integrated Computational Materials Engineering (ICME)



ICME Development Path

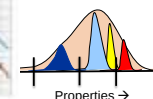
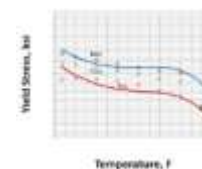
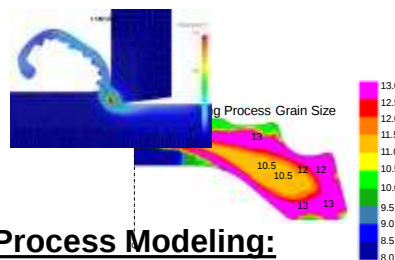
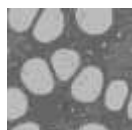
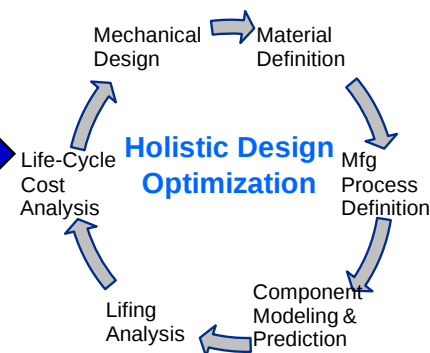
Capture benefits early and often



“Off-Ramps” for Developed Models
with FIT-FOR-PURPOSE Capabilities

Model
Development

Physic-based model application flow



Materials Modeling:

- Material Design / Optimization
- Materials Definitions / Design Curves

Mfg Process Modeling:

- Process Design / Optimization
- Design Rules and Tools

Component Modeling:

- Component Design / Optimization
- System Design Optimization

Interdisciplinary Applications

Local Applications



Global, “Holistic” Applications



Materials Data Perspectives



MIL-HBK-5H

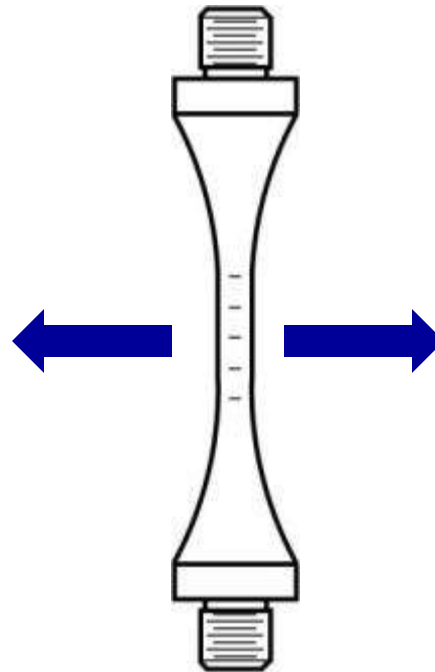
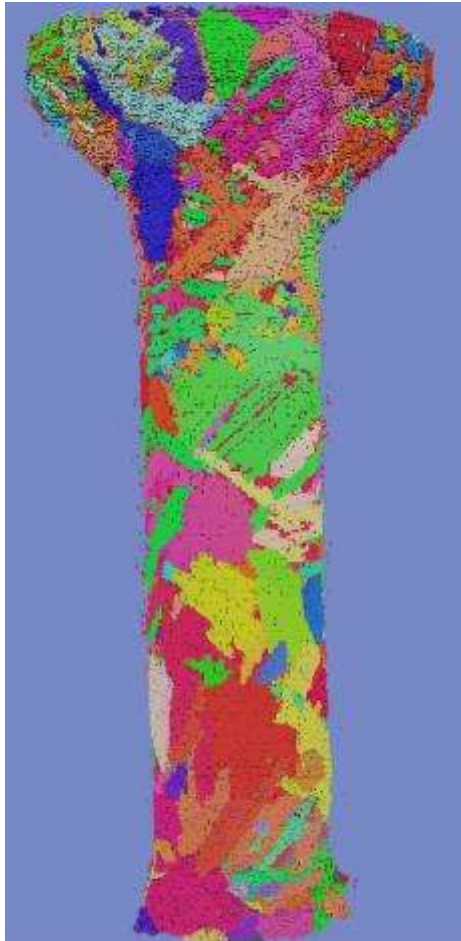


Table 5.4.1.0(h). Design Mechanical and Physical Properties of Ti-6Al-4V Sheet, Strip, and Plate

Specification	AMS 4911 and MIL-T-9046, Comp. AB-1					MIL-T-9046, Comp. AB-1				
Form	Sheet		Plate			Sheet, strip, and plate				
Condition	Annealed					Solution treated and aged				
Thickness, in.	0.1875		0.1875-2.000		2.001-4.000	0.1875	0.1875-0.750	0.751-1.000	1.001-2.000	
Basis	A	B	A	B	S	S	S	S	S	
Mechanical Properties:										
F_u , ksi:										
L	134	139	130 ^a	135	130	160	160	150	145	
LT	134	139	130 ^a	138	130	160	160	150	145	
F_y , ksi:										
L	126	131	120	125	120	145	145	140	135	
LT	126	131	120 ^a	131	120	145	145	140	135	
$F_{0.2}$, ksi:										
L	133	138	124	129	124	154	150	145	—	
LT	135	141	130	142	130	162	—	—	—	
$F_{0.01}$, ksi:	87	90	79	84	79	100	93	87	—	
$F_{0.005}$, ksi:										
(e/D = 1.5)	213 ^b	221 ^b	206 ^b	214 ^b	206 ^b	236	248	233	—	
(e/D = 2.0)	272 ^b	283 ^b	260 ^b	276 ^b	260 ^b	286	308	289	—	
$F_{0.002}$, ksi:										
(e/D = 1.5)	171 ^b	178 ^b	164 ^b	179 ^b	164 ^b	210	216	203	—	
(e/D = 2.0)	208 ^b	217 ^b	194 ^b	212 ^b	194 ^b	232	243	235	—	
α , percent (S-basis):										
L	8 ^c	—	10	—	10	3 ^d	8	6	6	
LT	8 ^c	—	10	—	10	3 ^d	8	6	6	
E , 10 ³ ksi	16.0									
E_r , 10 ³ ksi	16.4									
G , 10 ³ ksi	6.2									
ν	0.31									
Physical Properties:										
ρ , lb/in. ³	0.160									
C , K , and α	See Figure 4.5.1.0									

a The rounded T_u values are higher than specification values as follows: $F_u(L) = 131$ ksi, $F_u(LT) = 132$ ksi, and $F_u(LT) = 123$ ksi.

b Bearing values are "dry pin" values per Section 1.4.7.1.

c 8%—0.025 to 0.063 in. and 10%—0.063 in. and above.

d 7%—0.030 in. and above, 4%—0.033 to 0.040 in. and 3%—0.032 in. and below.

Materials Engineer

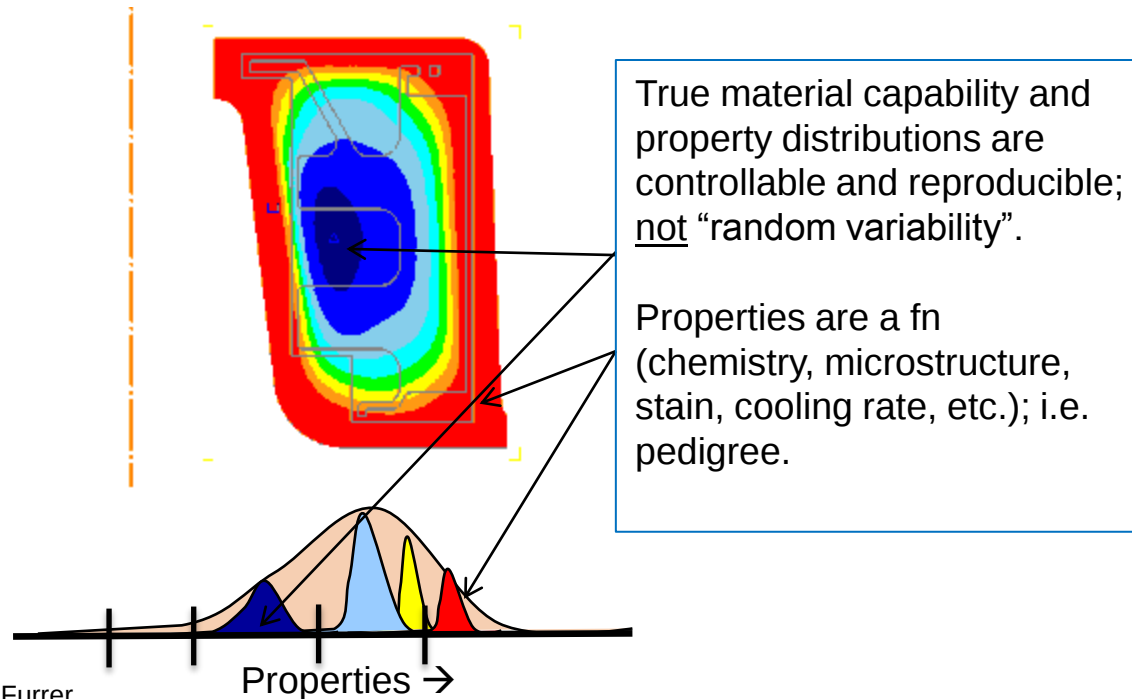
Mechanical Engineer



No Technical Data per the EAR or ITAR

2015 UCSB Workshop on Collection and Analysis of Big Data in 3D Materials Science

Material Capabilities Definitions



From D.Furrer,
OPTIMoM Conf., 2010

Materials properties are path dependent and are often “location-specific”. Engineering specifications often treat entire material volume as single, homogeneous property capabilities.

Modeling and simulation can help enhance component property capability definitions

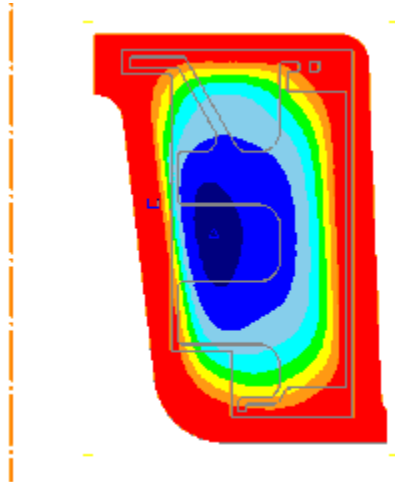


Efficient Testing and Qualification

Production Testing



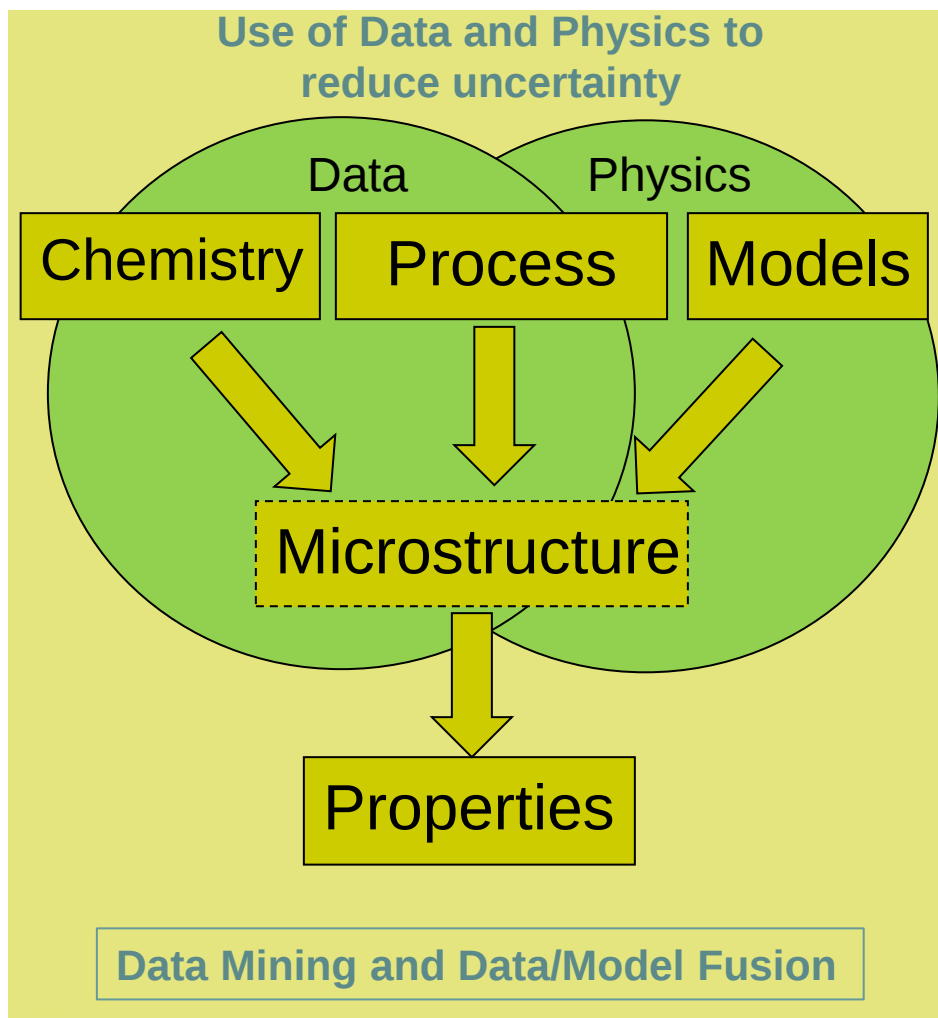
- Model-Driven Test Plans
 - Evaluation of critical component locations/properties
 - Target sensitive locations that provide information about entire component
 - Reduced testing provides reduced costs and qualification cycle-time



Critical – All Data to be Captured and Integrated with Current Understanding



Materials Informatics



- Capture and Re-Use Materials Data and Meta-Data (“Digital Thread”)
- Establish Enhanced Models to Support Future Materials Definitions and Design Functions
- Quantify Uncertainty of Models and Understanding to Minimize Future Testing



Materials Data Management

DEVELOPMENT DATA

Specific data collected, stored & reported



- **Material Data**

- Heat Code
- Material Spec
- Chemistry
- Source
- Part Number
- Heat Treatment
- Machining Source
- Material Source
- Expected Results



- **Administrative Data**

- Charge Numbers
- Purchase Order Approvers
- Export requirements
- Requestor
- Test Engineer
- Technical Contact
- Program Description

- **Tracking Data (barcode)**

- Approvals & Time Stamp
- Location & Time Stamp
- Alerts to incoming work
- Identify bottlenecks
- Locate specimen
- Job Status
- Performance assessment

- **Test Data**

- Test Conditions
 - Temp, Stress, Frequency, etc
- Test Results
 - Life, Strength, Curves (da/dN vs dK , S-S, etc.)

Microstructure Results



No Technical Data per the EAR or ITAR

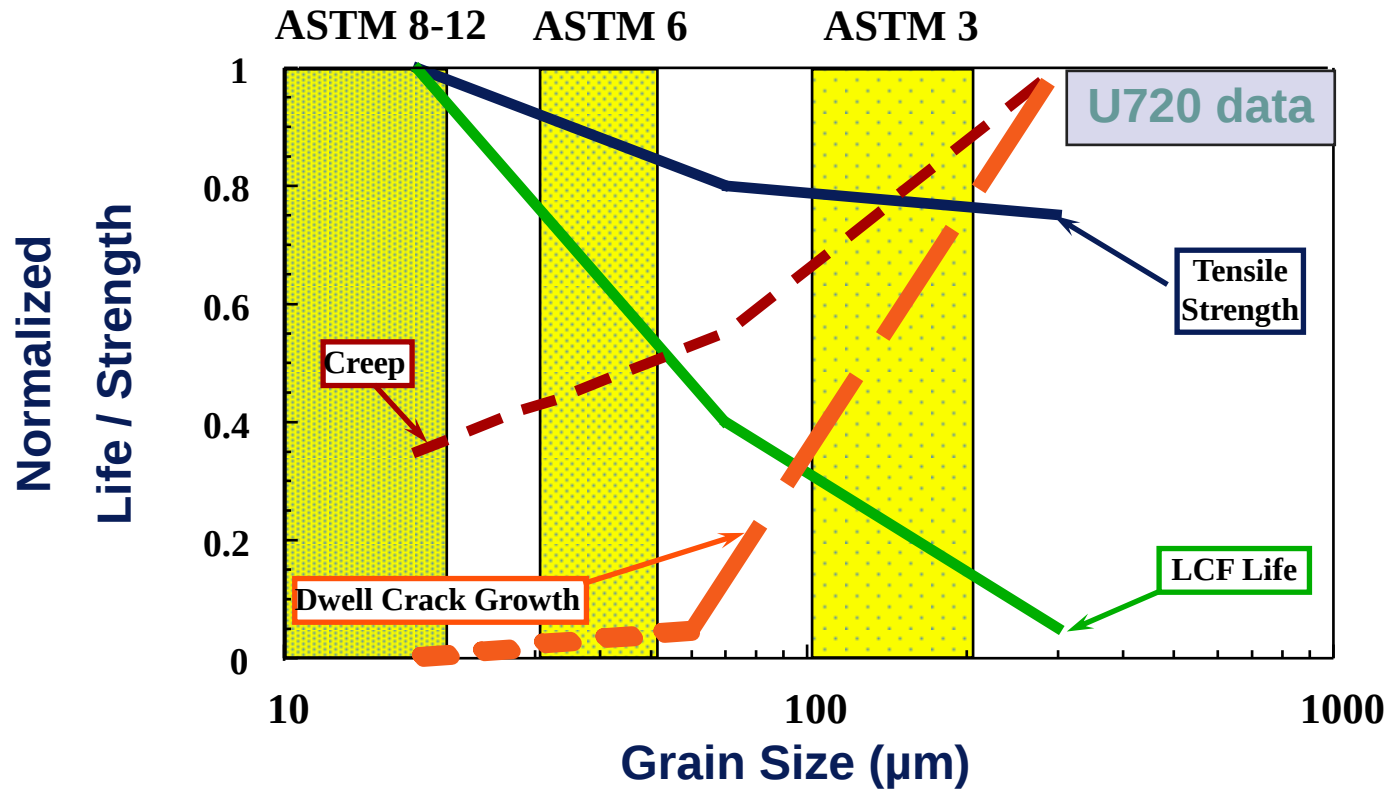
Measurement of Microstructure



- Quantitative metallography is critical to enable understanding influence of limiting features
- Distribution of features and tails of distributions provide keys to material behavior
- Historical method to quantify microstructure are no longer acceptable; greater statistical treatment is needed (SVEs, etc.)



Grain Size Impacts are Powerful



Significant implications for processing choices

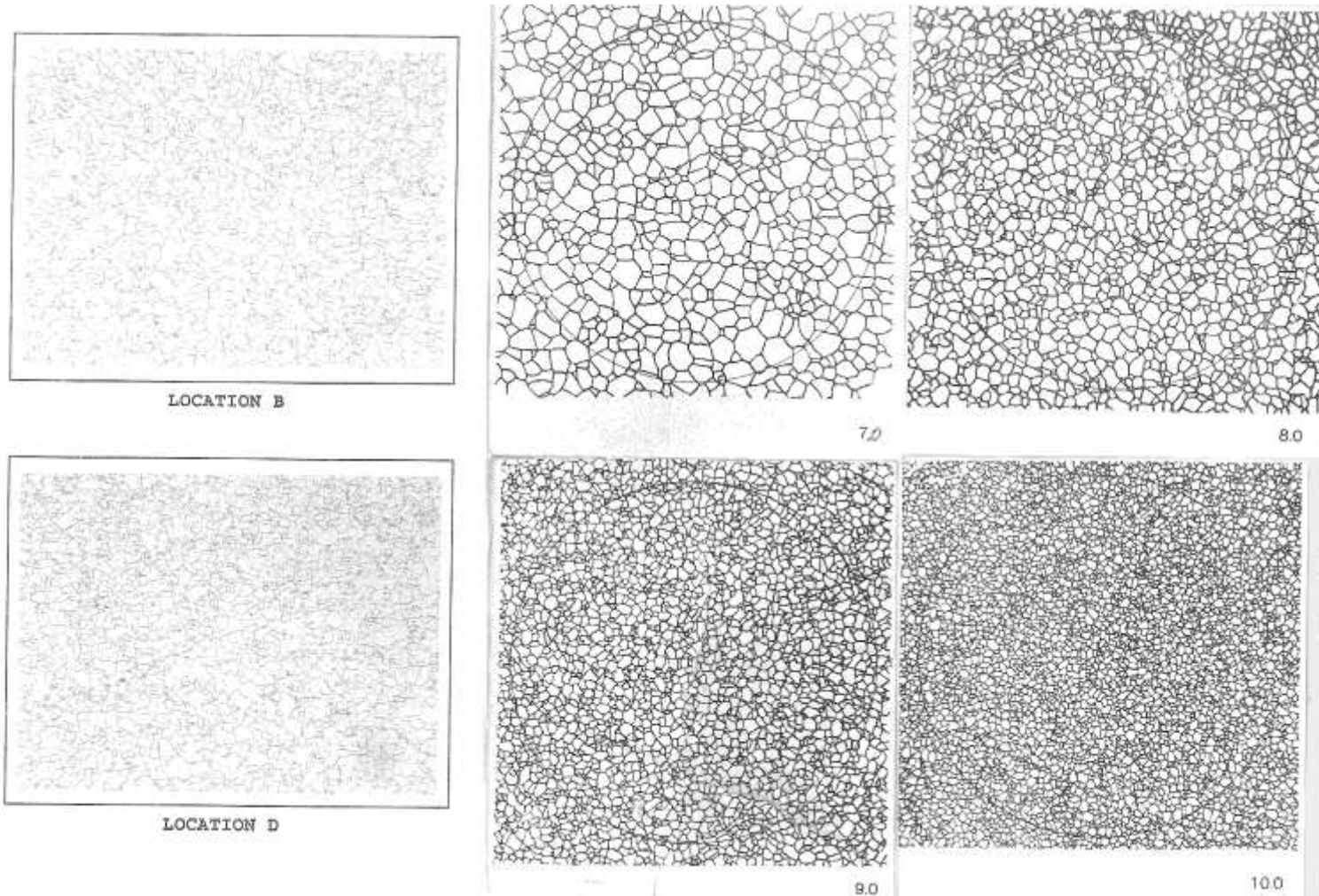
Source: J.Williams, OSU



No Technical Data per the EAR or ITAR

2015 UCSB Workshop on Collection and Analysis of Big Data in 3D Materials Science

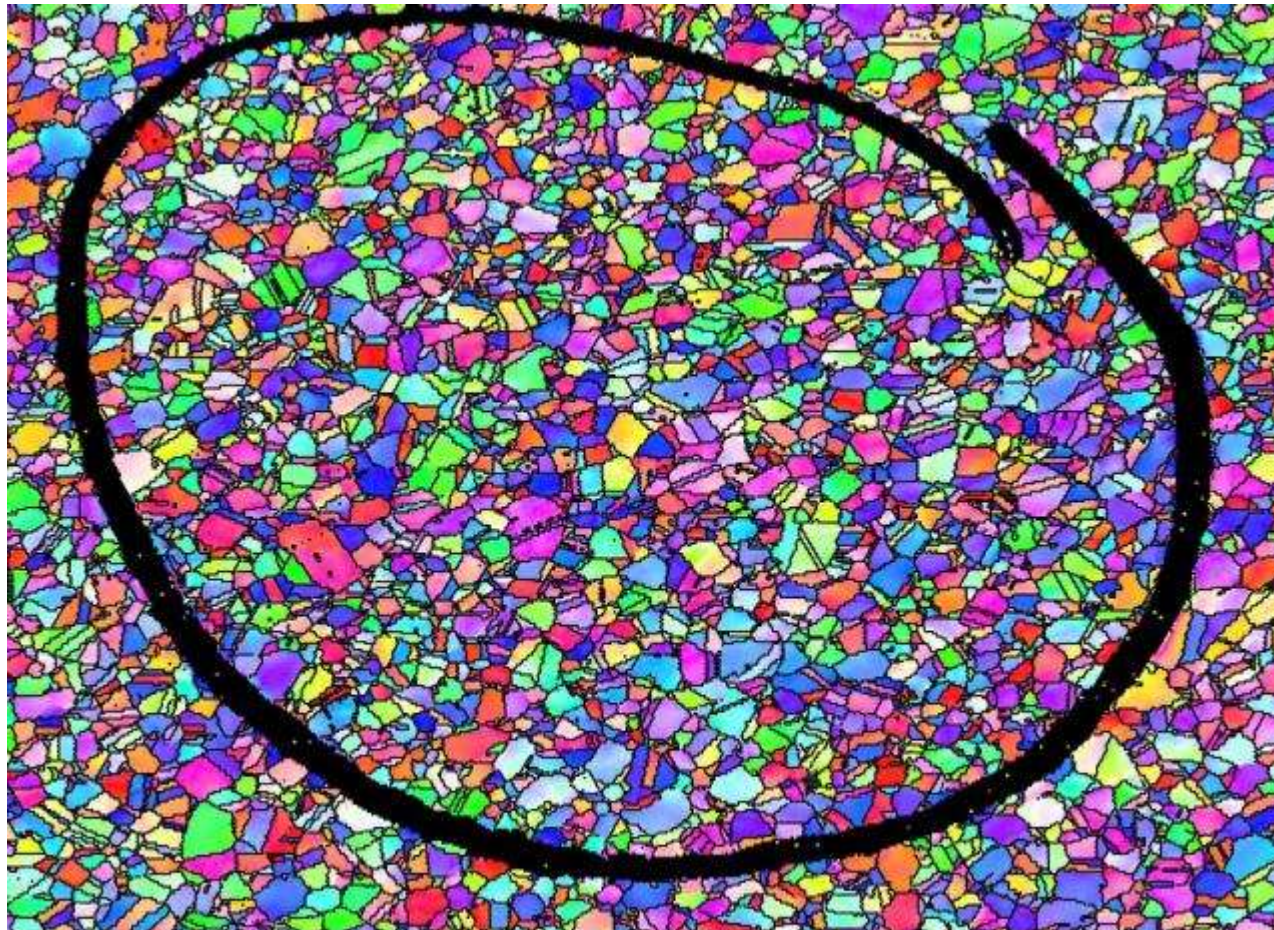
ASTM E112 Grain Size By Comparison Method



No Technical Data per the EAR or ITAR

2015 UCSB Workshop on Collection and Analysis of Big Data in 3D Materials Science

Grain Size Distribution in Waspaloy Material



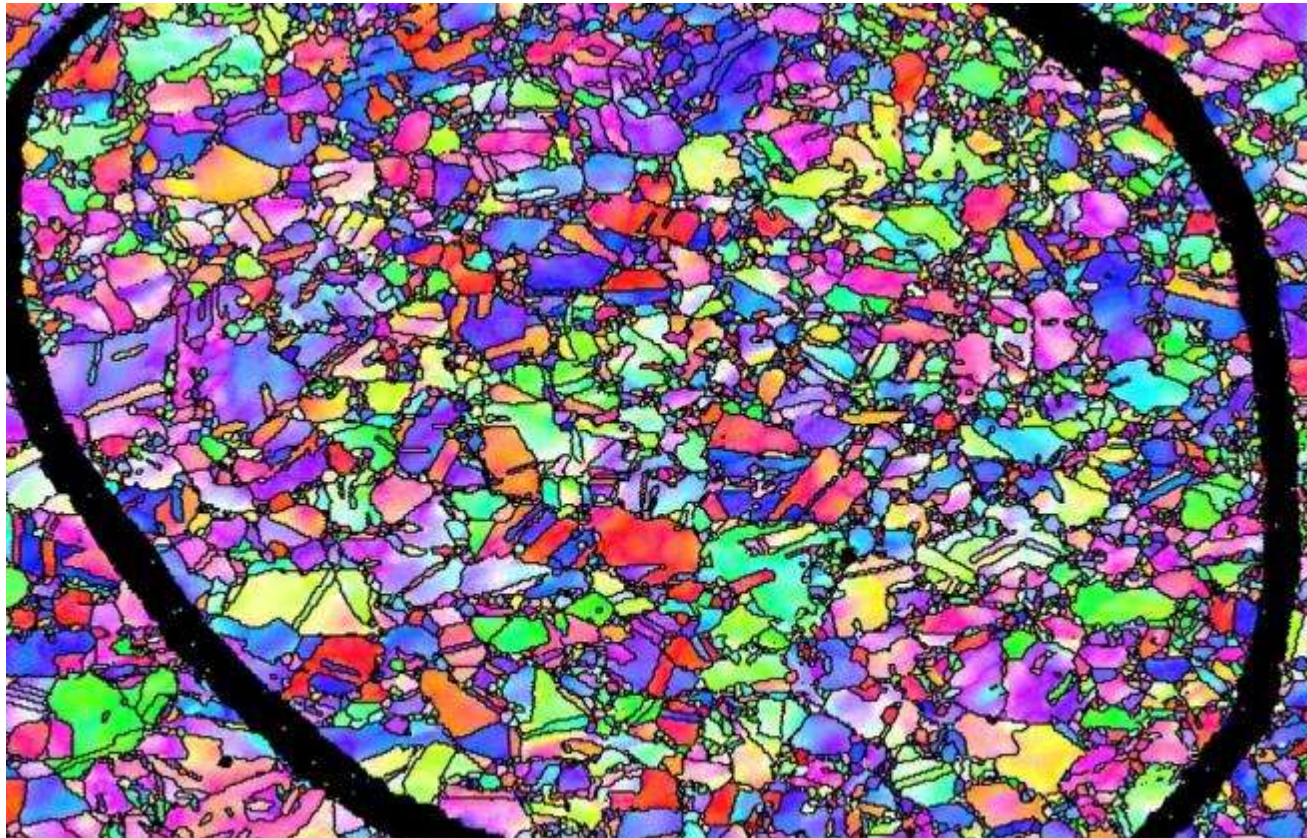
700.00 μm = 100.0 steps
Boundary levels: 15°
IPF [001]



No Technical Data per the EAR or ITAR

2015 UCSB Workshop on Collection and Analysis of Big Data in 3D Materials Science

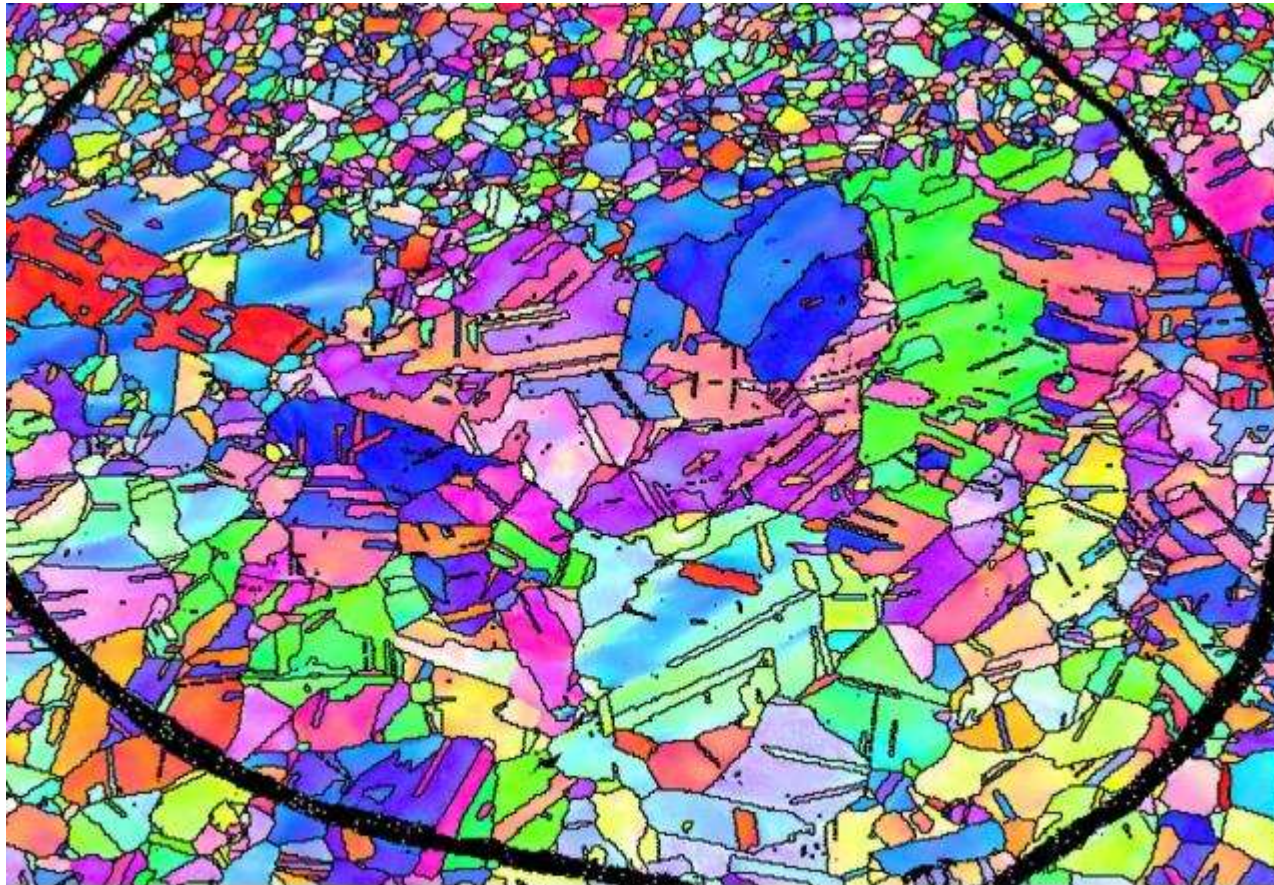
Necklaced Grain Structures in Waspaloy Material



500.00 μm = 100.0 steps Boundary levels: 15°
IPF [001]



Non-Uniform Grain Size in Waspaloy Material



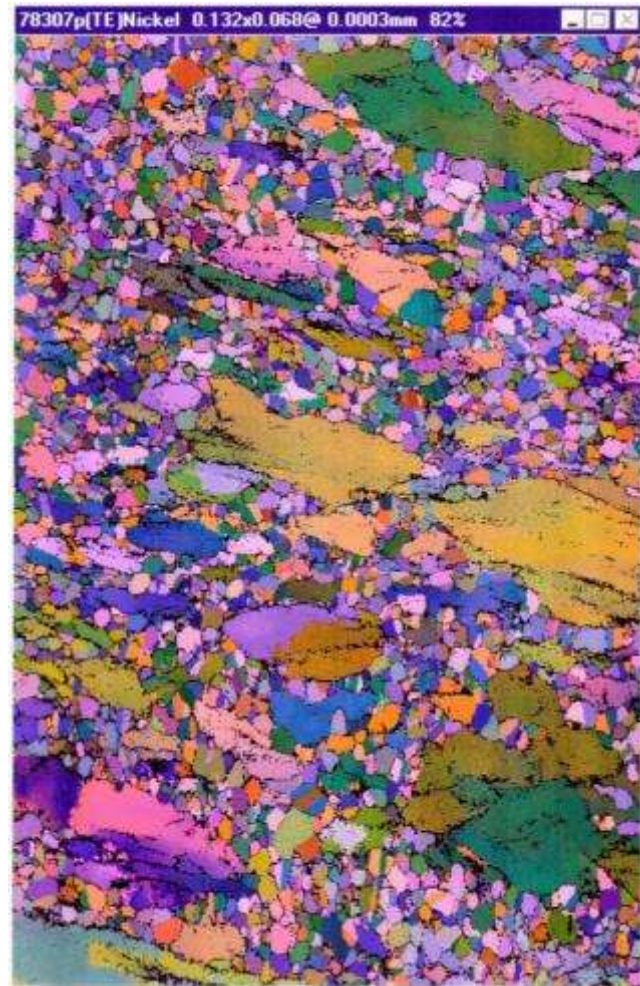
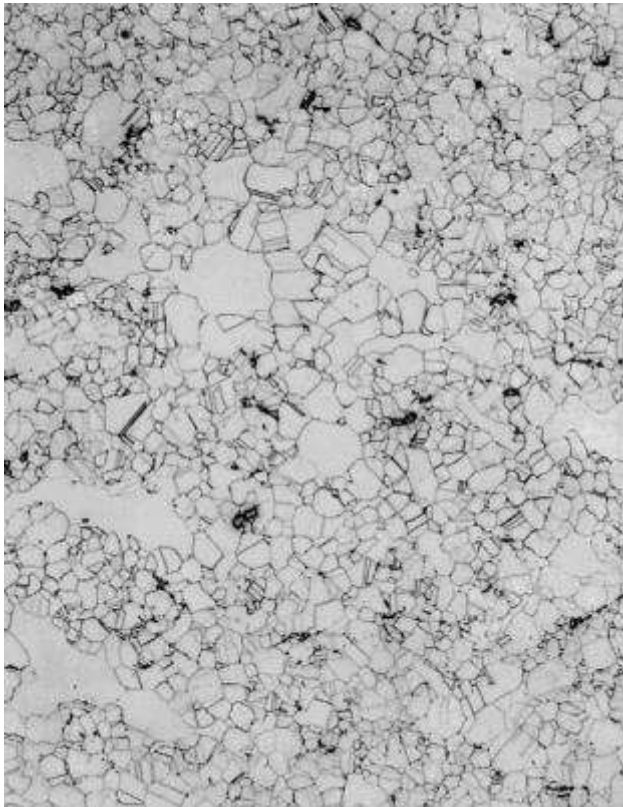
700.00 μm = 100.0 steps
Boundary levels: 15°
IPF [001]



No Technical Data per the EAR or ITAR

2015 UCSB Workshop on Collection and Analysis of Big Data in 3D Materials Science

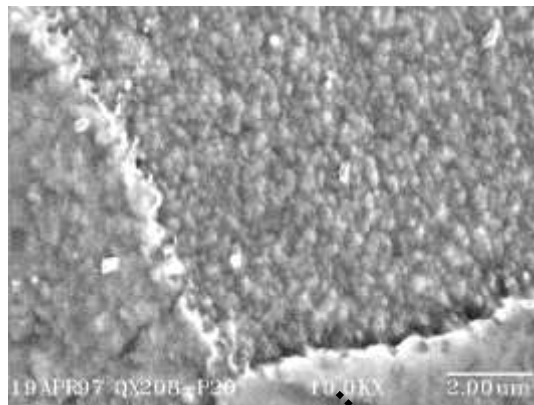
Optical and OIM Images of Partially Recrystallized Wapaloy



No Technical Data per the EAR or ITAR

2015 UCSB Workshop on Collection and Analysis of Big Data in 3D Materials Science

Grain Boundary Morphology in Superalloy U720

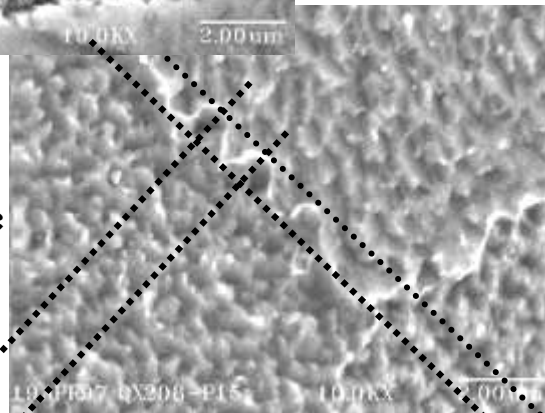


12.7°C/sec



0.12°C/sec

1.3°C/sec



0.72°C/sec*

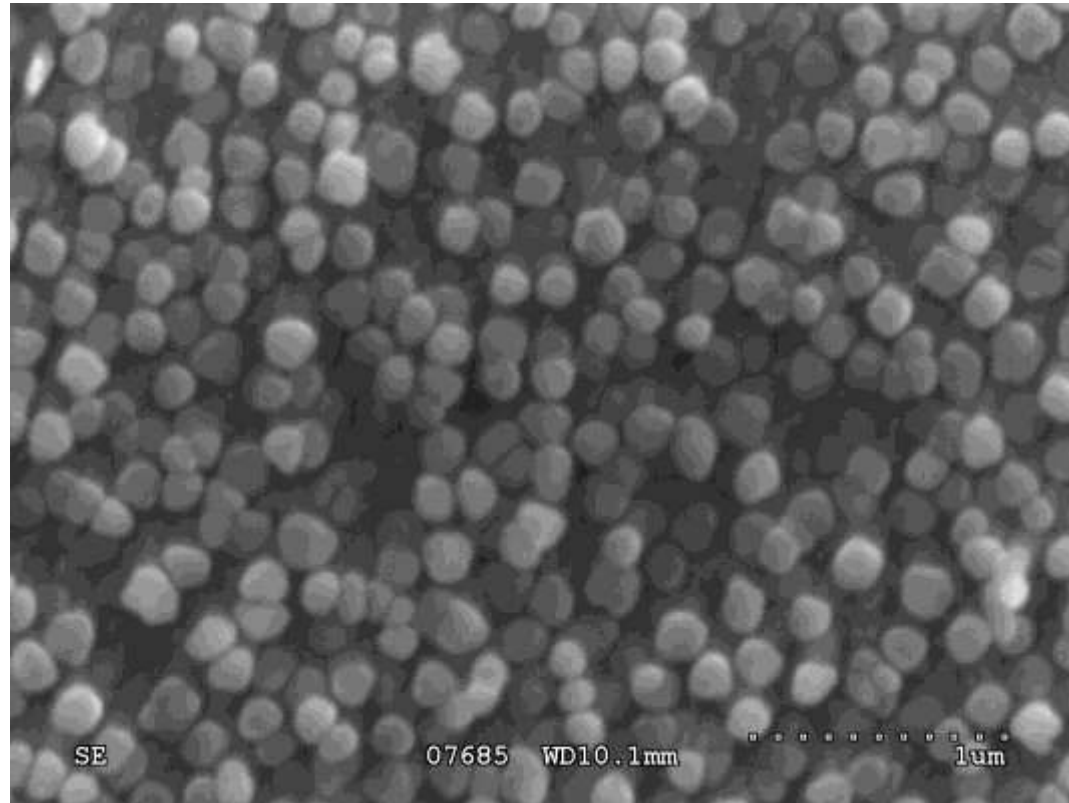


Amplitude

Periodicity



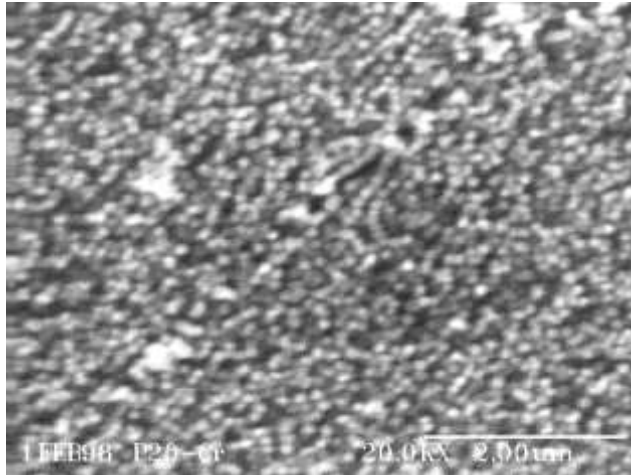
Gamma-Prime Size and Distribution in P/M Superalloy



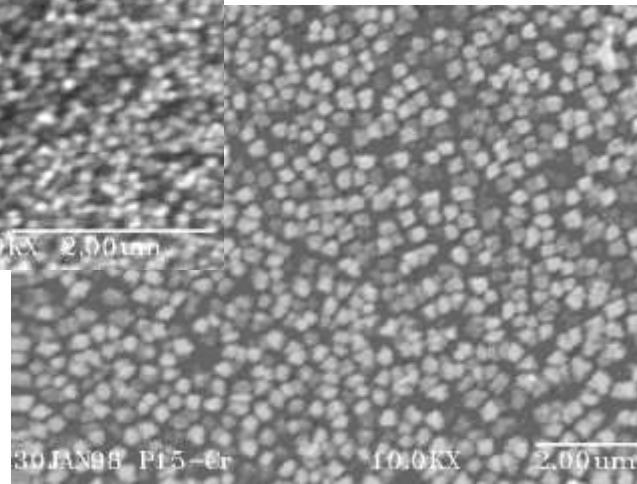
No Technical Data per the EAR or ITAR

2015 UCSB Workshop on Collection and Analysis of Big Data in 3D Materials Science

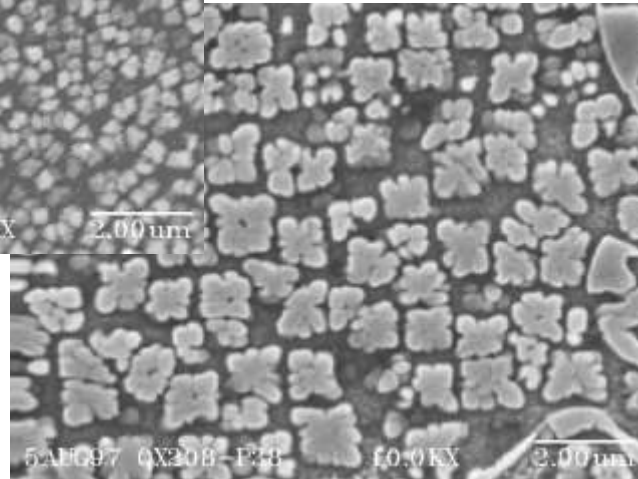
Gamma-Prime Precipitate Size as a Function of Cooling Rate



12.7°C/sec



1.3°C/sec



0.12°C/sec

Gamma-Prime Morphology

Gamma-Prime Density

Gamma-Prime Size

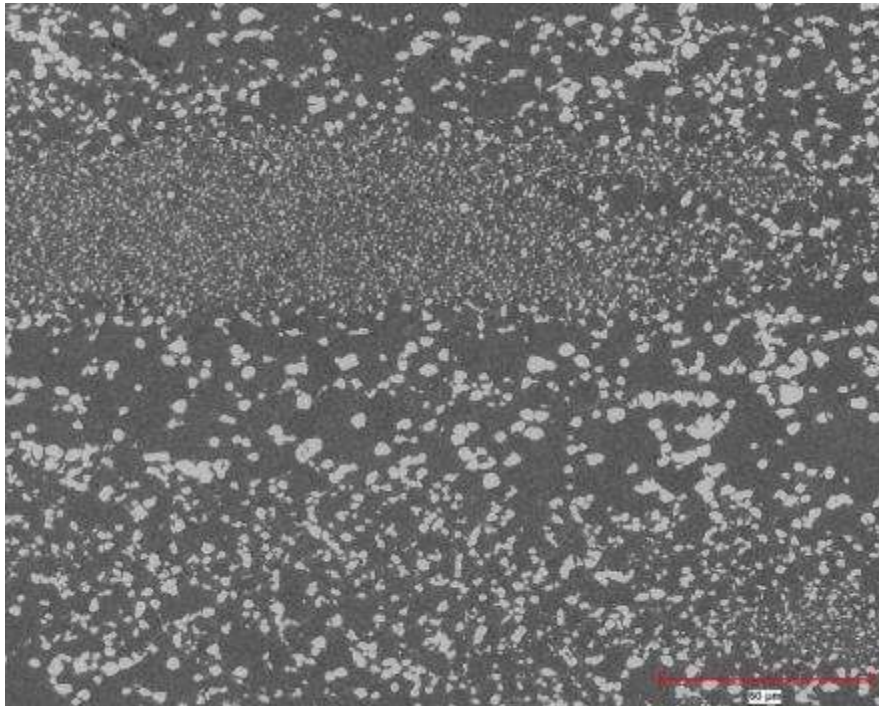
U720



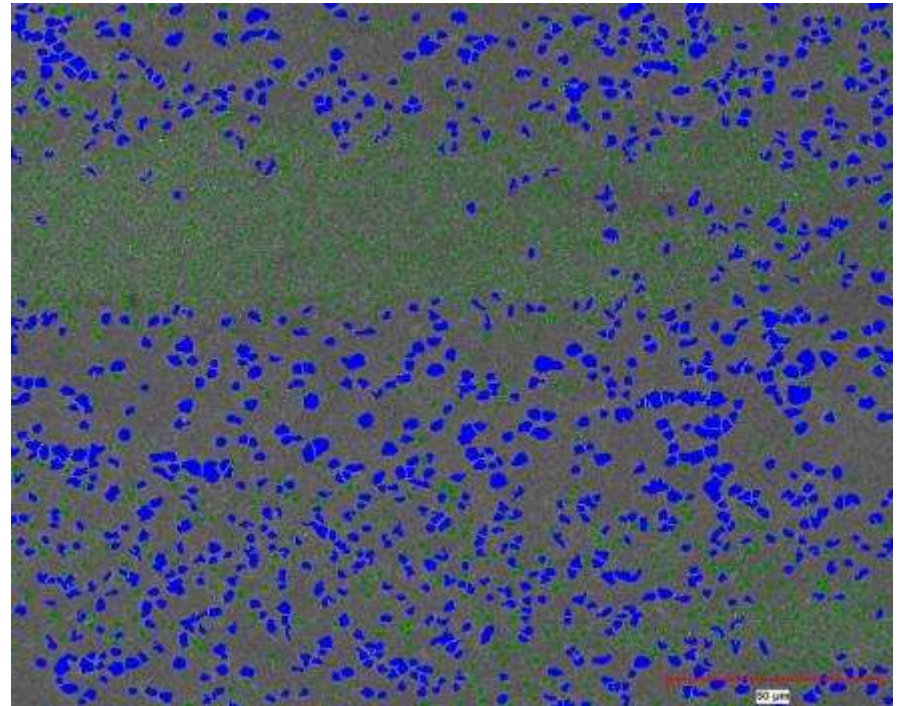
No Technical Data per the EAR or ITAR

2015 UCSB Workshop on Collection and Analysis of Big Data in 3D Materials Science

Gamma-Prime Banding in Superalloy U720



Area Percent of Large Grains 12.7%
Area Percent of Small Grains 6.43



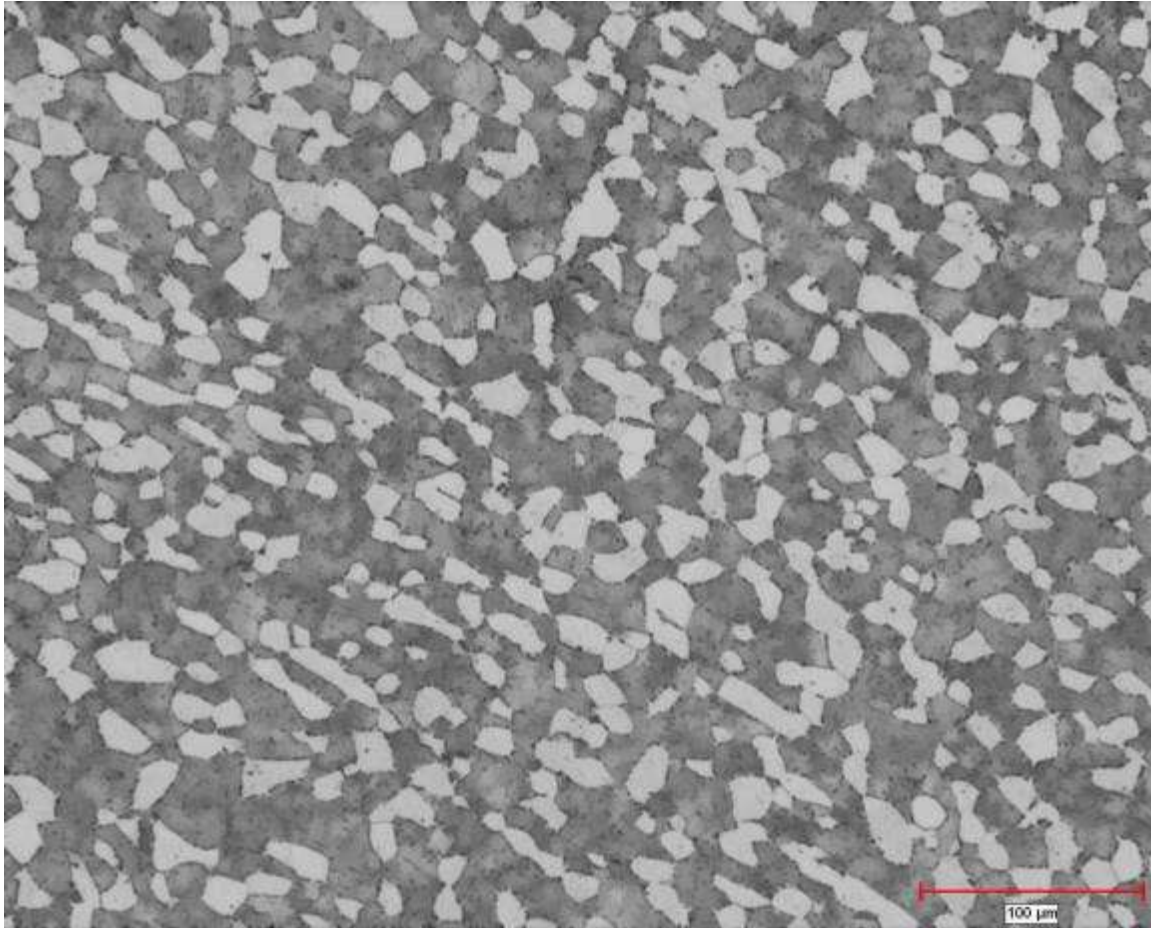
Average Size of Large Grains $5.25 \mu\text{m}^2$
Average Size of Small Grains $0.626 \mu\text{m}^2$



No Technical Data per the EAR or ITAR

2015 UCSB Workshop on Collection and Analysis of Big Data in 3D Materials Science

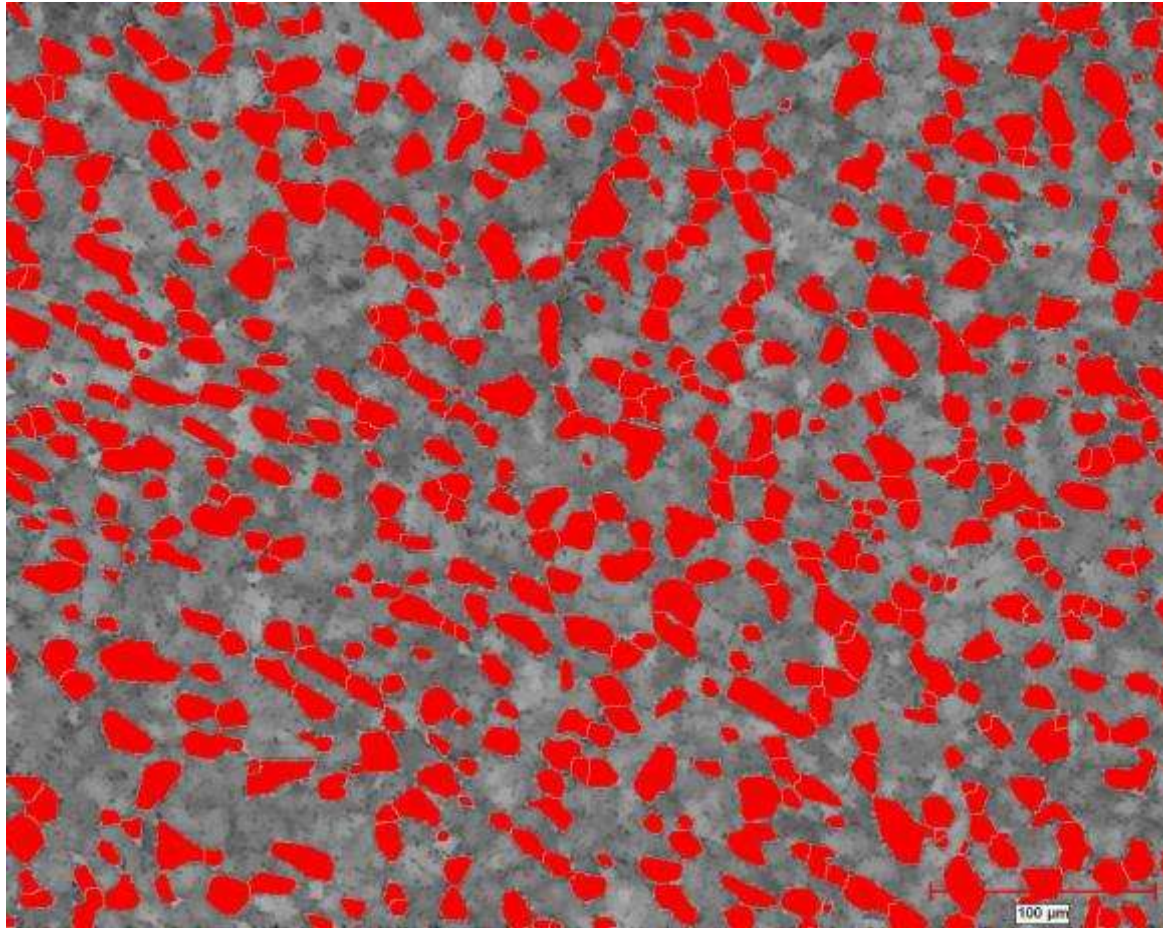
Primary Alpha Size and Volume Fraction in Titanium



No Technical Data per the EAR or ITAR

2015 UCSB Workshop on Collection and Analysis of Big Data in 3D Materials Science

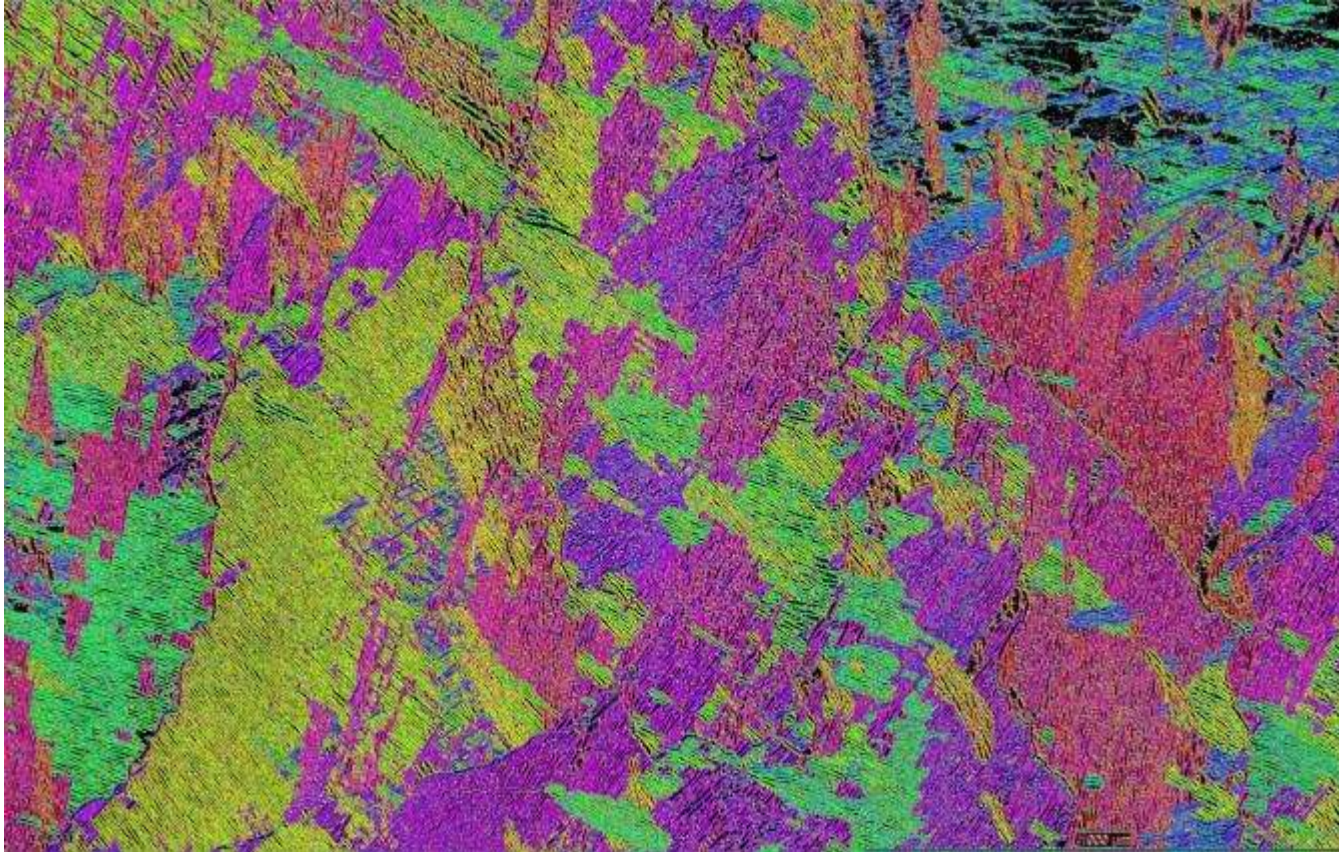
Primary Alpha Size and Volume Fraction in Titanium



No Technical Data per the EAR or ITAR

2015 UCSB Workshop on Collection and Analysis of Big Data in 3D Materials Science

Optical Image Analysis of Ti Transformed Beta Structure



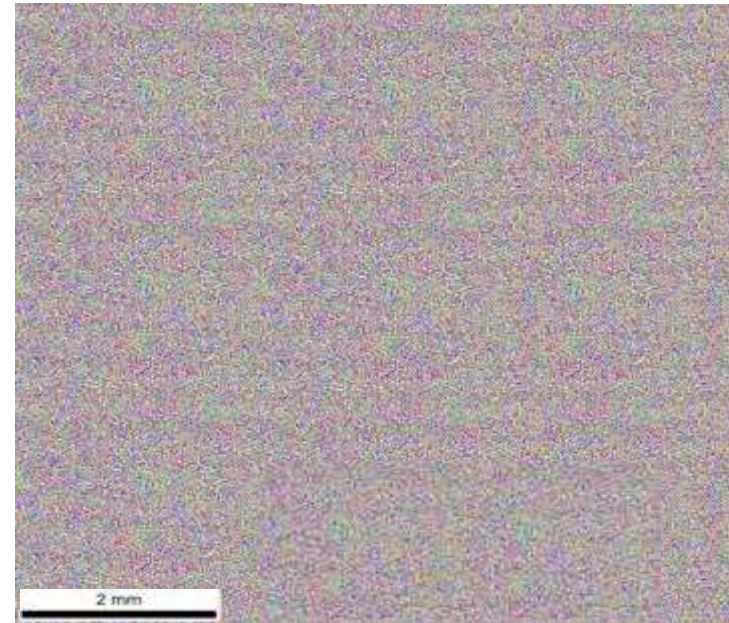
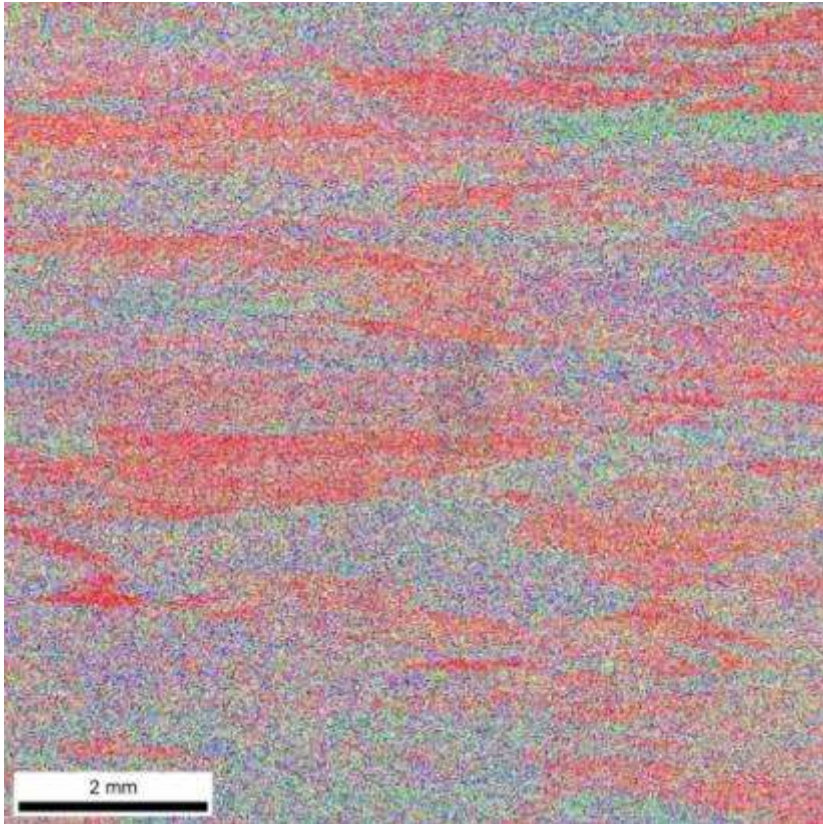
M. Dallair and D. Furrer, "Quantitative Metallography of Titanium Alloys",
Advanced Materials and Processes, December, 2004, pp. 25-28.



No Technical Data per the EAR or ITAR

2015 UCSB Workshop on Collection and Analysis of Big Data in 3D Materials Science

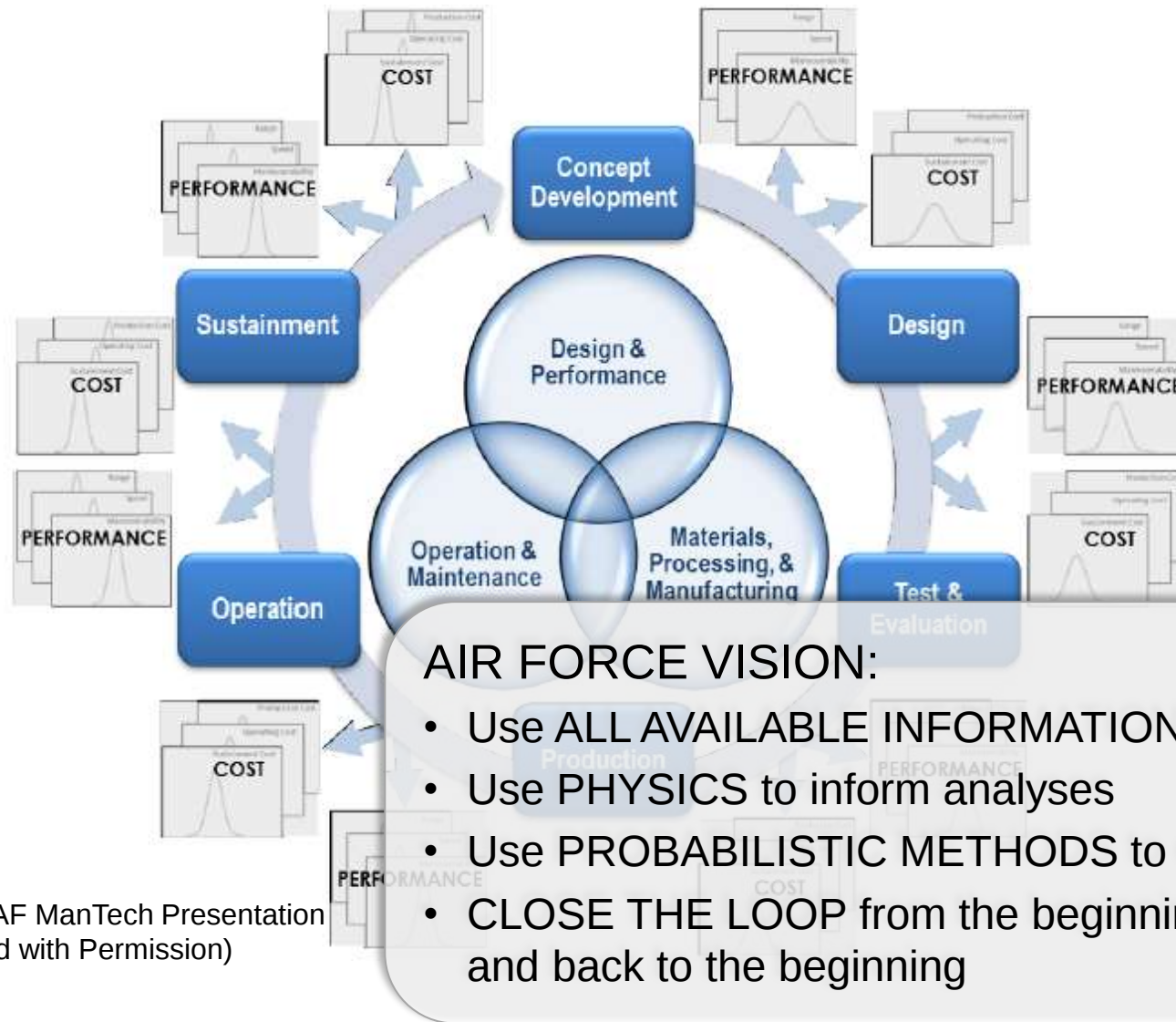
Micro-Texture in Titanium Alloys



No Technical Data per the EAR or ITAR

2015 UCSB Workshop on Collection and Analysis of Big Data in 3D Materials Science

Digital Thread - Air Force Vision



Source: USAF ManTech Presentation
(Used with Permission)



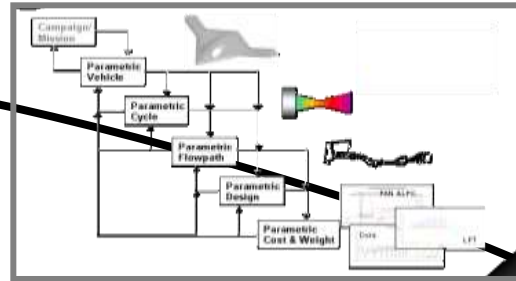
Turbine Disk Digital Thread



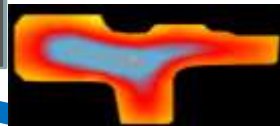
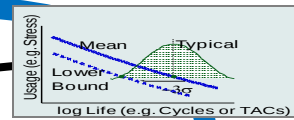
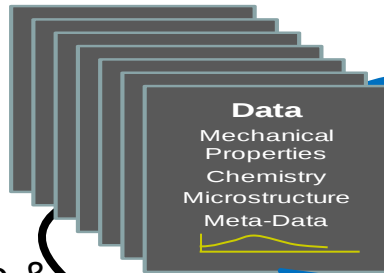
ICMSE - Integrated Computational Materials
Science & Engineering



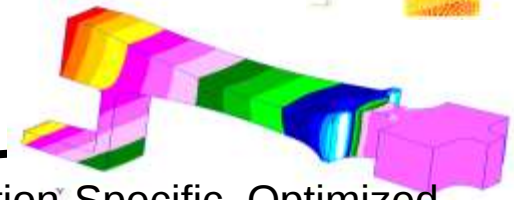
System Requirements



PMDO – Preliminary
Multi-Disciplinary
Optimization



Materials Genome Initiative



Location-Specific Optimized
Component Definition

Microstructure &
Process-Sensitive
Materials Definition



Data Capture /
Knowledge Generation



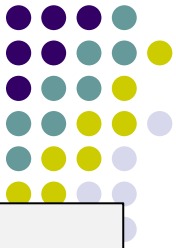
No Technical Data per the EAR or ITAR



Model-Based
Manufacturing Definition

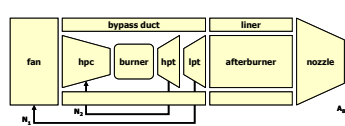


Digital Thread & Design for Variation Vision



Digital Thread Enables More Effective Decision Making Through DFV

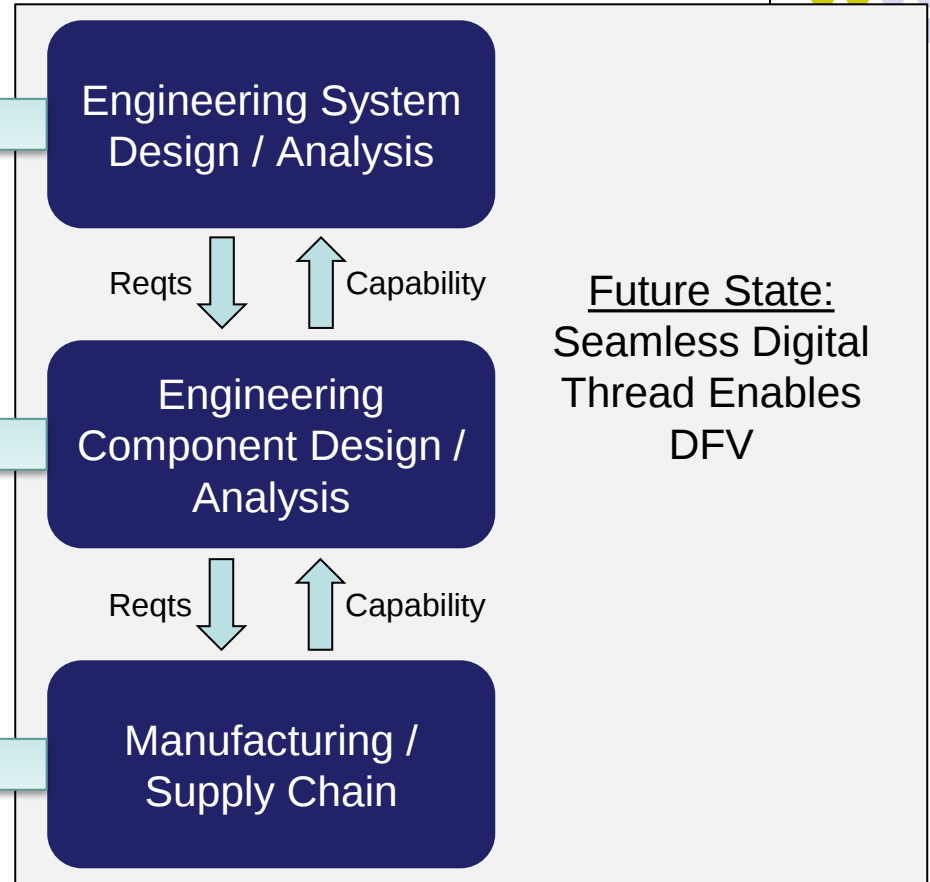
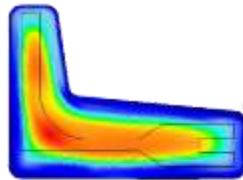
System Engineering Models



Multi-Disciplinary Automated Workflows



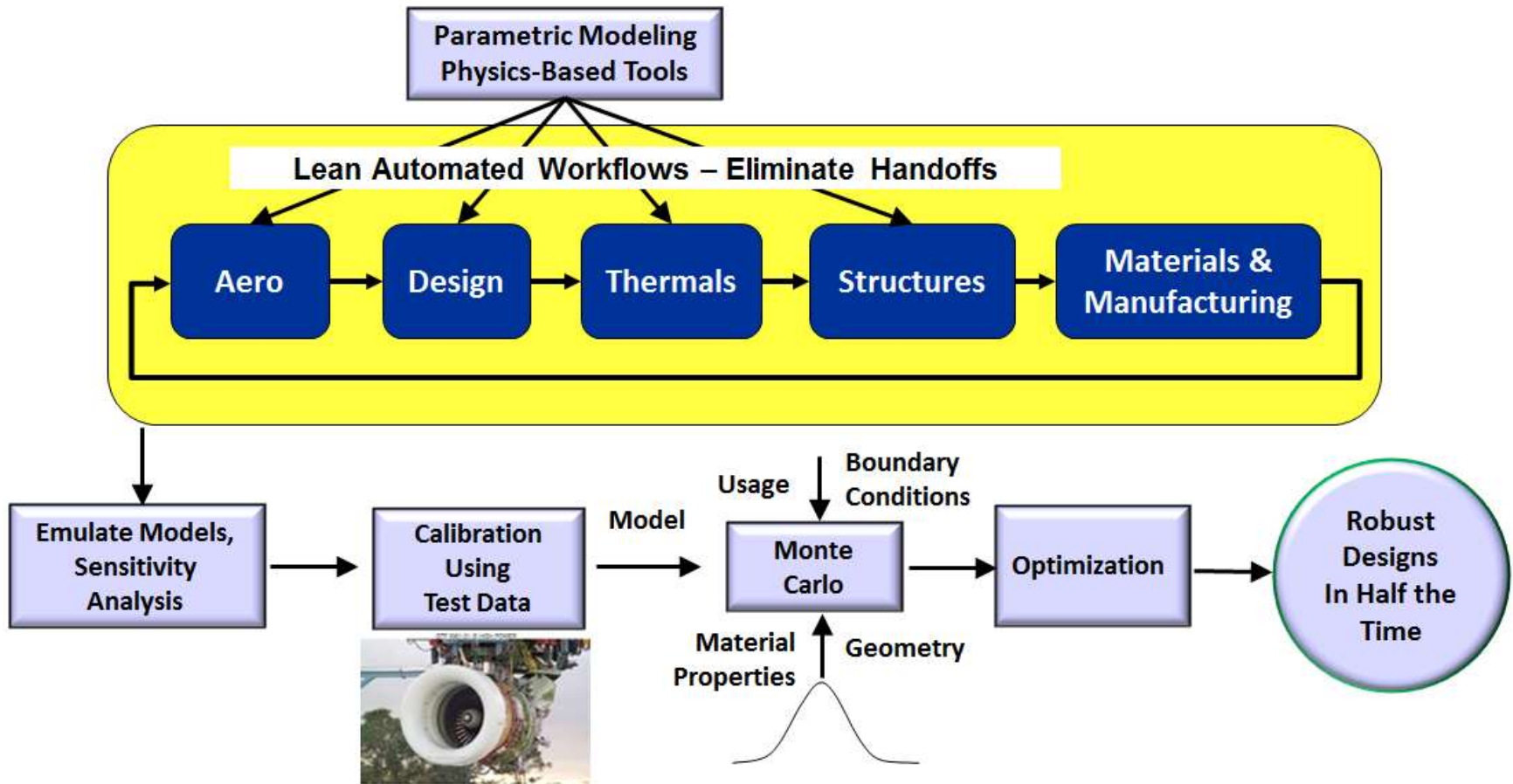
Materials and Manufacturing Process Modeling



Goal is 50% reduction in development time with 90% elimination of quality non-conformance



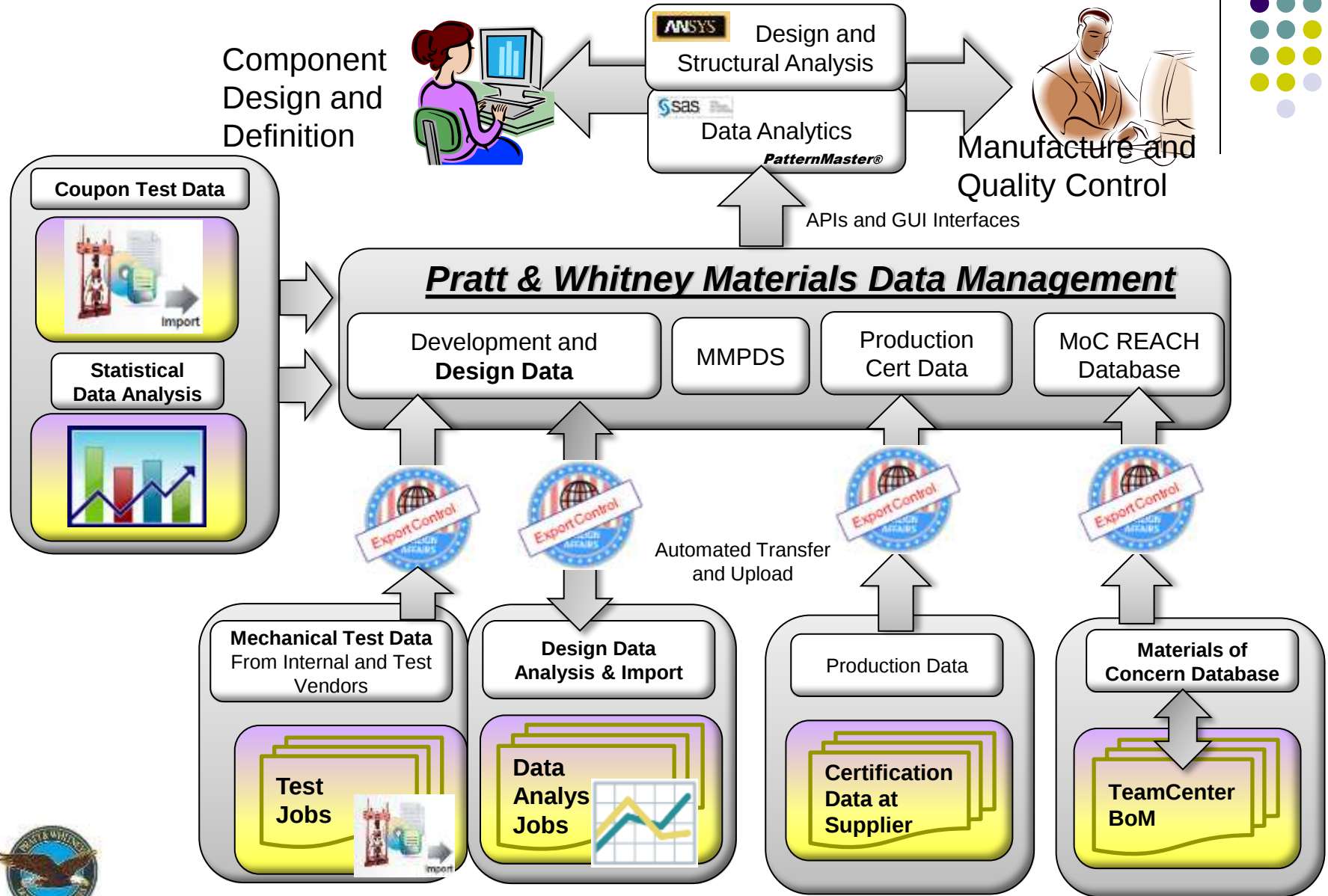
Design for Variation



DFV Enables Reduced Development & Part Costs by Aligning Models, Data, and Producibility

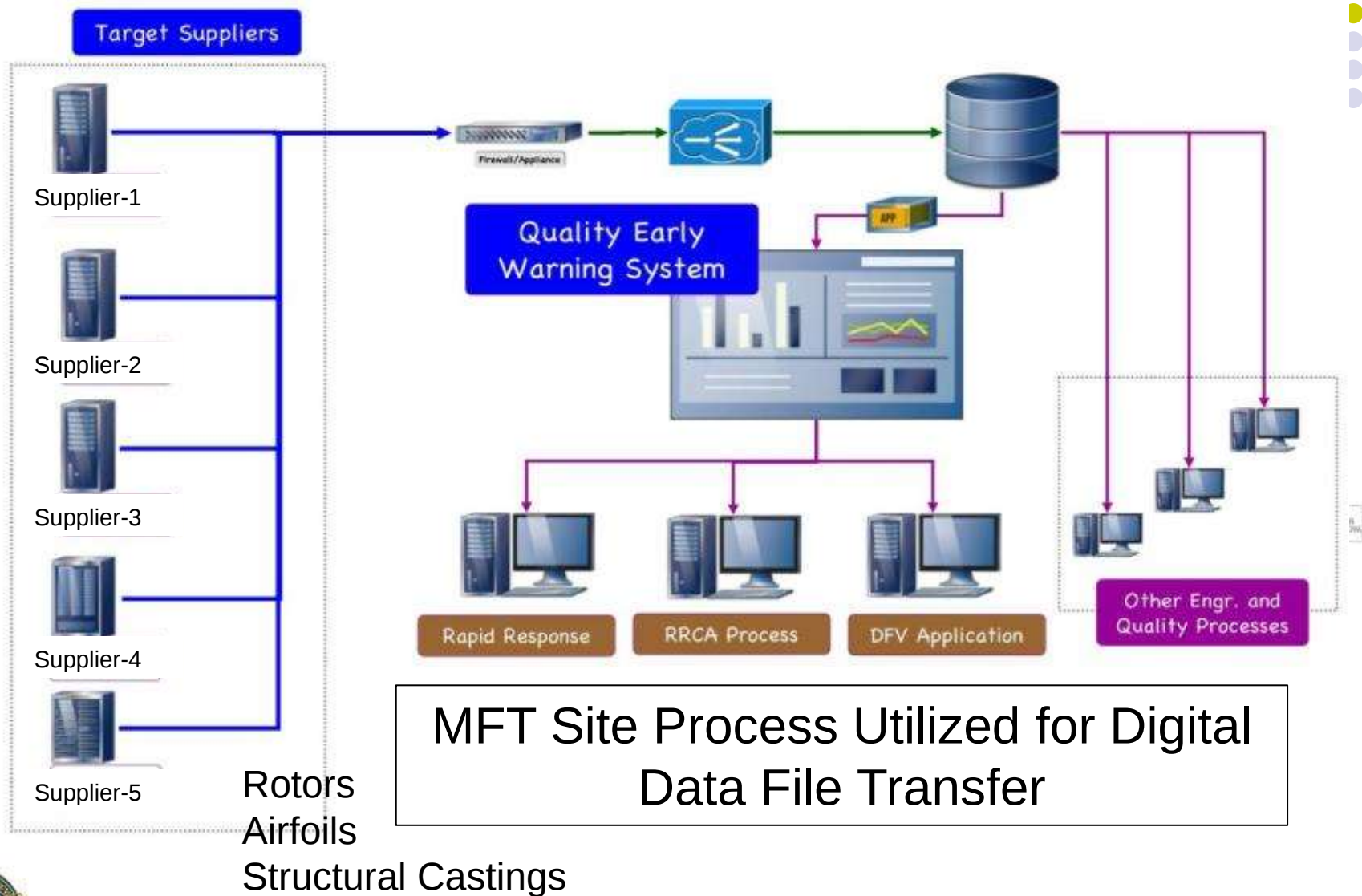


Materials Data Management



No Technical Data per the EAR or ITAR

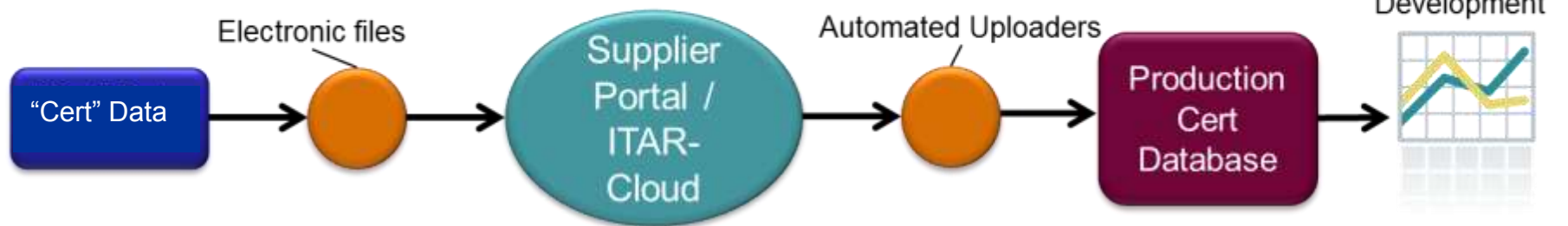
Materials Data Management



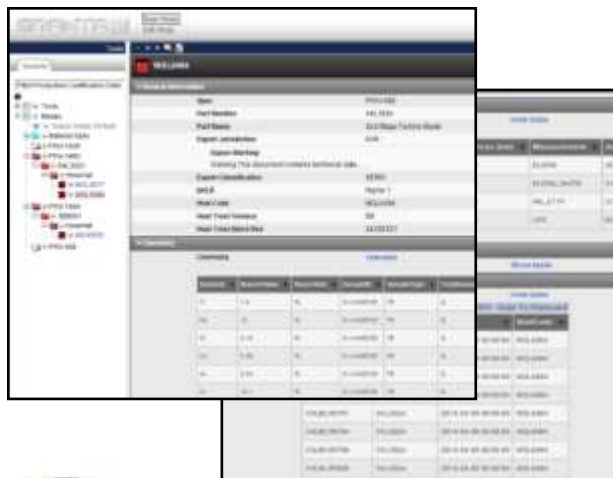
Materials Data Management Initiative

Data Capture Efforts showing Success

Data for over hundreds of thousands of serials captured to-date



GRANTA-MI is Live



- Production Data Pilot Proving Digital Data Capture Is Possible
- Data has been Analyzed and Applied to Support and Establish Root Cause on Two Recent Materials QNs



- Production data has been captured, analyzed, linked to models and applied to enhanced material definitions
 - Castings, Forgings
- Optimal chemistry range defined for several legacy alloys, which supports required manufacturing control

A scatter plot showing the relationship between Rupture Life at 18000psi (hrs) on the x-axis and Rupture Life at 18000psi (hrs) on the y-axis. The data points are scattered, showing a positive correlation. A red regression line is drawn through the data points, and a red horizontal line is drawn at y=45.



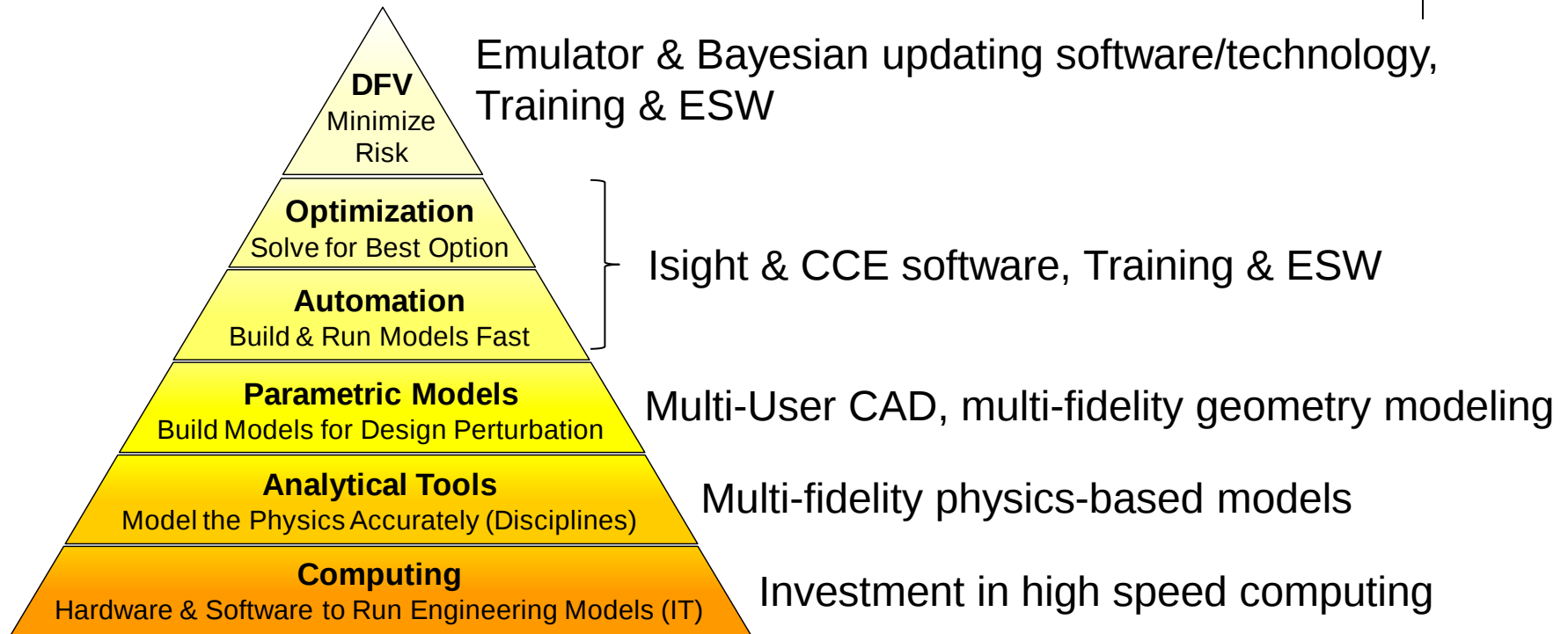
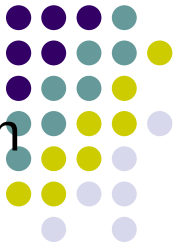
The screenshot shows the NetLogo interface with the 'Behavioral' tab selected. On the left, there are three monitors for 'AI_Sentences', 'AI_Costs', and 'AI_Resources'. In the center, there is a table titled 'Behavioral Data' with columns for 'AI_Sentences', 'AI_Costs', and 'AI_Resources'. The table shows data for 10 different scenarios, with values ranging from 0 to 1000. The 'AI_Sentences' column shows values like 1000, 1000, 1000, 1000, 1000, 1000, 1000, 1000, 1000, 1000. The 'AI_Costs' column shows values like 1000, 1000, 1000, 1000, 1000, 1000, 1000, 1000, 1000, 1000. The 'AI_Resources' column shows values like 1000, 1000, 1000, 1000, 1000, 1000, 1000, 1000, 1000, 1000.



No Technical Data per the EAR or ITAR

Design for Variation

Enabling Multi-Fidelity and Multi-Disciplinary Uncertainty Quantification



Internal Investment in Key Enablers Driving Broad Implementation of DFV



No Technical Data per the EAR or ITAR

2015 UCSB Workshop on Collection and Analysis of Big Data in 3D Materials Science

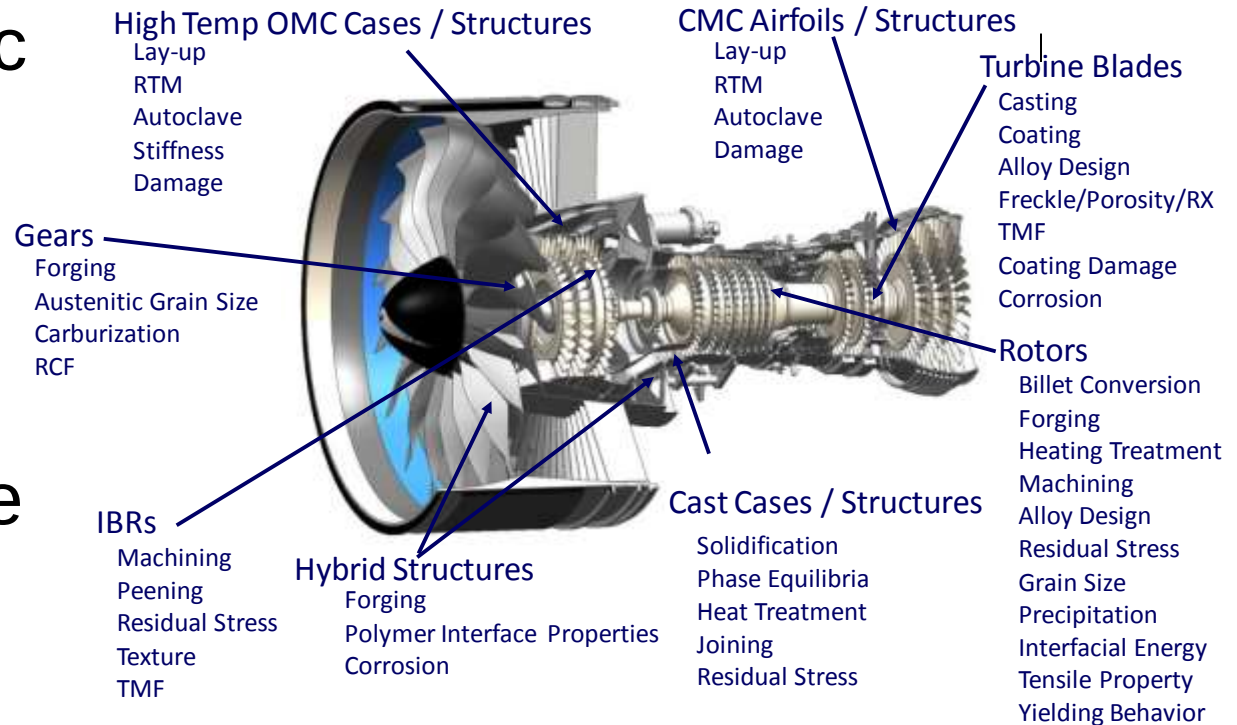
Computational Materials Modeling



Enhance specific materials capabilities

Design for cost and manufacture

Rapid, robust materials definitions



Many Applications for Materials and Process Modeling within Turbine Engines





Conclusions

- Prediction of material and component properties is a major element of ICME
- Understanding of weakest microstructural links and associated mechanisms is required
- Traditional characterization methods not acceptable for understanding limiting features relative to statistics and tails of distributions
- Appropriate length scales of SVEs required based on feature and mechanism
- “Big Data” and data analytics are critical elements of ICME

