

Acoustic Signatures and Machine Learning in

Additive Manufacturing

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Introduction and Motivation



- Laser Powder Bed Fusion can print complex high-quality parts.
- EOS M290 has over 100 process parameters to control.
- Fast cooling rate (10⁶ K/s) creates fine microstructure.
- LPBF has a handful of identified defect signatures.





a) T. DebRoy et al., Prog. Mater. Sci. 92 (2018) 112–224.
b) R. Li, et al., Int. J. Adv. Manuf. Technol. 59 (2012) 1025–1035.
c) W.E. King, et al., J. Mater. Process. Technol. 214 (2014) 2915–2925.
d) X. Yan, et al., Mater. Sci. Eng. A. 789 (2020) 139615.



Introduction and Motivation



Signature	Frequency (kHz)	Material	Detection Method	Source
Crack	100	Al6Mn2Ce	Structure- borne Acoustics	Seleznev et. al., 2022
Conduction Mode	0-20	316L SS, Bronze, Inconel 718	Airborne Acoustics	Drissi-Daoudi, et. al., 2022
Lack of Fusion	0-20	316L SS, Bronze, Inconel 718	Airborne Acoustics	Drissi-Daoudi, et. al., 2022
Keyhole	43.8	Ti6Al4V	Optical (X-ray)	Khairallah, et. al., 2021
Machine Noise	0.5	304 SS	Airborne	Ye, et. al., 2018
Material Resonance	1.16	304 SS	Airborne	Ye, et. al., 2018







Introduction and Motivation



Processing-Structure Relationship



Mission Relevance



- Laser powder bed fusion has been shown to have a higher risk to nuclear nonproliferation than other additive manufacturing methods.
- CoCrFeMnNi HEA has shown resistance to void swelling.





N. Cannon, et al., J Radioanal Nucl Chem 331,

Lu, C., et al., Nat Commun 7, 13564 (2016).





High Throughput Experiments



242 hex nuts specimens:

- Different processing conditions
- High throughput characterization
 - Density and Hardness Measured
- Rapid identification of processing bounds







High Strain Rate Experiments



High strain rate experiments conducted at Los Alamos National Laboratory measured the spall strength of the CoCrFeMnNi HEA across seven processing conditions.





V.K. Euser, et. al., Materialia, 33 (2024)



High Strain Rate Experiments





Spall response dominated by presence of cracks.

V.K. Euser, et. al., Materialia, 33 (2024)





High Strain Rate Experiments



Ultimate Tensile Strength changes by 4%

Yield Strength changes by 10%

Spall Strength changes by 30%







Acoustic Signatures









Acoustic Signatures









- Random Forest Model from scikit-learn package using 20 trees with no bootstrapping.
- 50% of data used for training. Each processing condition is represented in each dataset.
- Chebyshev polynomials (15 terms) applied to FFT to reduce data size.
- The predictions from 5 clips of each sample are averaged to get the final prediction for each samples.

Machine Learning



Random Forest Simplified

Instance

Tree







Technical Work and Results





The Random Forest Algorithm can differentiate high defect

parts with 82% accuracy.





Technical Work and Results





The Random Forest Algorithm can differentiate between high and low spall strength with no errors.





Future Experiments









Conclusion



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Conclusion



Processing-Structure Relationship



ETI Impact and Future Work



- What is the impact of the ETI on your development?
 - This work was presented at MS&T 2023 in Columbus, OH.
- Personnel transitions:
 - I will be conducting a summer internship at LLNL in Summer 2024 studying the physical mechanisms of acoustic signatures in LPBF.
- Technology transitions:
 - The plate impact experiments were conducted in conjunction with Saryu Fensin and Ginny Euser at Los Alamos National Laboratory.
- Future Work:
 - Future experiments will explore acoustic signatures up to 100 kHz to look for signatures of discrete cracking events.
 - Acoustic signatures of other materials will be analyzed to determine material dependent acoustic signatures.





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