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Machine learning assisted material development via Laser powder bed fusion process

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Introduction

Materials and Methods

Results and Discussion

Conclusions

Thesis Outcomes





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Introduction



- Producing components from a digital CAD model
- Adding material layer by layer







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- ✓ Design freedom
- ✓ Material efficiency
- \checkmark Customization



Open challenges:

▲ Poor surface quality



- ▲ Limited available materials
- ▲ Low production rate

How can we turn these challenges into future strengths?

Introduction

Challenges and solutions



Materials and Methods

Powder Preparation

Jar milling (97.5%AISI316L+2.5%Cu)

Laser Powder Bed fusion

Mechanism explanation: A laser beam is used to **selectively melt** fine metal **powder** and build up fully-dense parts **layer-by-layer** Laser Scanner

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Materials and Methods

Design Of Experiment: 64 different process parameters

Power
 100, 190, 200, 270, 340 W

Scan Speed
 400, 600, 800, 1000 mm/s

- ➢ Hatch Distance 0.1, 0.11, 0.12, 0.13, 0.2 mm
- Layer Thickness0.03 mm



Materials and Methods

DESIGN OF EXPERIMENTS: 64 Cubic Specimens



Powder Feedstock: AISI316L-2.5%Cu



Metal AM Technology: L-PBF



Analyzed conditions: As-Built



Characterizations: Powder, Density Process Parameter Optimization,







Applying Machine Learning models



Mean Absolute Error

Average absolute deviation for each statistic
The same unit as the target variable
The lower, the better!







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Coefficient of Determination

$$-\infty < R^2 < +1$$

Level of variation in the target variable by changes in the input variables

R² =
$$\frac{\sum_{i=1}^{n} (\hat{y}_i - \bar{y})^2}{\sum_{i=1}^{n} (y_i - \bar{y})^2}$$

 $MAE = \frac{1}{n} \sum_{i=1}^{n} |y_i - \hat{y}_i|$ $R^2 = \frac{\sum_{i=1}^{n} (\hat{y}_i - \bar{y})^2}{\sum_{i=1}^{n} (y_i - \bar{y})^2}$





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Support Vector Regression







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gamma





R² or MAE alone can not be considered for

- comparison of prediction accuracy.
- Although the training size of 90% has the highest R², its MAE value is high.
- The **best algorithm** was **SVR**, with a training size of 80%.



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Model Verification(Comparison of actual and predicted density using SVR at 80% training size based on VED)



N. point	Actual Relative Density (%)	Predicted Relative Density (%)	Error (%)
1	90.90	92.21	1.45
2	95.00	95.03	0.04
3	97.84	97.78	0.06
4	99.15	98.93	0.22
5	98.87	99.06	0.19
6	98.52	99.05	0.53
7	98.89	97.71	1.20
8	98.46	98.07	0.40
9	97.81	98.19	0.39
10	96.44	96.64	0.21
11	97.88	96.22	1.69
12	98.58	97.48	1.12
13	97.38	97.28	0.10
14	97.05	97.42	0.38
15	97.74	96.45	1.32





Process parameters defect content relationship



Conclusions

- 1. ML techniques accurately predicted the relationship between process parameters and defect content and enhanced the building rate, thus addressing quality and productivity issues in the L-PBF process.
- The SVR model (training size: 80%) was the best model for predicting the relationship between 2. process parameters and defect content.

The SVR model achieved a 0.62% average error, demonstrating its reliability for optimizing 3. process parameters in L-PBF and minimizing material waste.







Any Question!

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