Priv.-Doz. Dr. Stefan Bosse, Dr. Dirk Lehmhus

How can we detect and characterize hidden defects in materials low-cost and high-quality? Reducing defects in materials improves robustness, ecology, and economics.

OBJECTIVES

Robust detection and characterisation of hidden defects, e.g., pores or cracks, is a still challenge.

Knowledge about defects enables improvement of design, production, and life-time.

Fraunhofer **DFG** FOR3022





Improving Ecology and Economics of High-pressure Aluminum Die Casting Processes A data-driven analytical characterization of hidden pores and defects using low-cost X-ray radiography images and advanced simulation

- Common Measuring Techniques: X-ray Radiography/CT, US Sonography Visual inspection and characterization is error prone and time-consuming Automated damage and defect diagnosis is required > But data-driven models require **ground**
 - truth data set and advanced work flow



METHODS

- > X-ray Radiography (LowQ, MidQ)
- > X-ray Computer Tomography

- > X-ray Image Simulation
- > CAD Modeling of defects using **Monte**
 - **Carlo Simulation**
- Semantic CNN Pixel Classifier and **DBSCAN** for Feature Marking

Reconstruction (Reference, HighQ)



Stefan Bosse Universität Bremen Mathematik & Informatik Universität Siegen Maschinenbau



Dirk Lehmhus Fraunhofer IFAM Materialwissenschaften Bremen

Priv.-Doz. Dr. Stefan Bosse, Dr. Dirk Lehmhus

The Reality Gap: Noise, Bias, and Artifacts Good and Bad News **Do not trust data-driven models!**

RESULTS

A simple data-driven feature marking detector is suitable to detect pores in **Iow-contrast and Iow-cost X-ray** radiography images.

Simulation of X-ray image data sets for the training of the detector is valid and suitable.



Fraunhofer **DFG** FOR3022



Improving Ecology and Economics of High-pressure Aluminum Die Casting Processes A data-driven analytical characterization of hidden pores and defects using low-cost X-ray radiography images and advanced simulation

✓ A data-driven image feature detector trained using synthetic data only can be applied to real measured images □ But noise, simulation bias, and computational artifacts decrease the feature marking accuracy (too much FP) ✓ Missing Ground truth problem solved by simulation







CONCLUSION

- noise
- models



Noise: Gaussian detector and non-Gaussian X-ray

Pores=Defects (TP)

TP: True Positive (Pore) **FP: False Positive** FN: False Negative (Back.) **TN: True Negative**

> Semantic pixel classifier is robust against Gaussian detector noise, but highly sensitive to non-**Gaussian spatially correlated X-ray**

> Due to the missing ground truth in real world images, the feature marking model must be trained with synthetic images derived from CAD

Artifacts (FP) were observed in feature maps of synthetic X-ray images independent of the CAD model and of defect-free materials!



Stefan Bosse Universität Bremen Mathematik & Informatik Universität Siegen Maschinenbau



Dirk Lehmhus Fraunhofer IFAM Materialwissenschaften Bremen