

*Deep Learning for Quality  
Prediction in Metal  
Additive Manufacturing*

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# What is *Laser Based Additive Manufacturing*?

- Laser Based Additive Manufacturing (LBAM) is synonymous with layer manufacturing where products are made layer-by-layer
- LBAM can produce complicated parts with a wide breadth of engineering capabilities
- LBAM produces a melt pool from a laser
- The melt pool is used to produce the parts, layer-by-layer, and is also an integral part in determining the microstructure of the parts it creates



# *Problems in Additive Manufacturing*

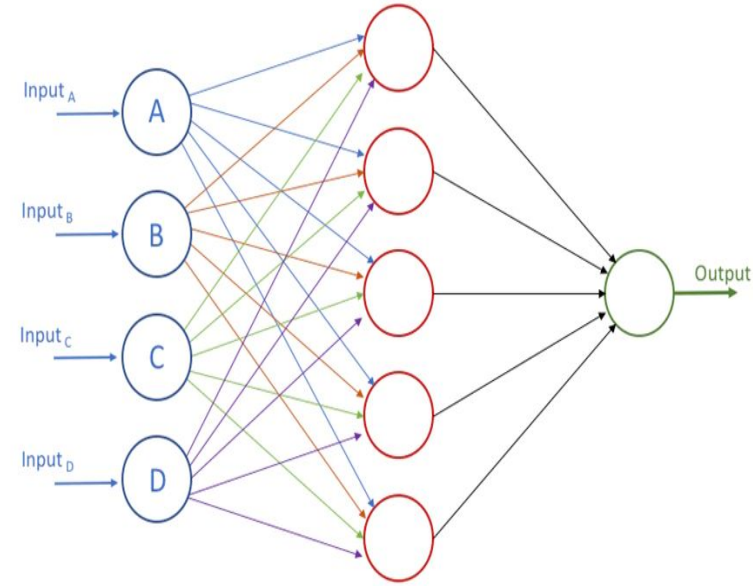
- A defect in LBAM is porosity, very small holes that occur in the surface
  - Occurs when the substrate is not fully melted/fused
- These defects can cause some structures to be unstable or adversely affect performance, this can be hard and expensive to detect

# *Current and Future Approaches*

- Using physical and statistical models to predict when such defects would occur
  - Many complex mechanical models are used involved to make this type of detection, which increases cost
  - Not realistic for mass-production nor real-time quality control
- Use machine learning along with real time sensors to detect anomalies

# What is Machine Learning?

- Machine Learning is a subset of Artificial Intelligence that uses data (typically quite a lot) with statistical algorithms to find patterns in data
- Neural Networks



# *Machine Learning + Additive Manufacturing*

- Use data collected from pyrometer (measuring melt pool temperatures) and an Infrared (IR) Camera (measuring IR energy and mapping to produce a thermal visualization of the melt pool).
- From the data construct a reliable and accurate neural network that can predict defects during the LBAM process

# *Questions to Explore*

- Possibilities of automatically cropping images in areas where defects are likely to occur
- How to incorporate an already existing physics knowledge into a machine learning model
- Finding what data is most valuable/influential in prediction of defects

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