

Artificial intelligence and data-driven material and device engineering

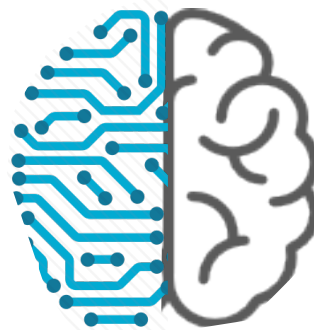
materials science data

100100101
010010010
101110111
011100110

training



AI model



predicting



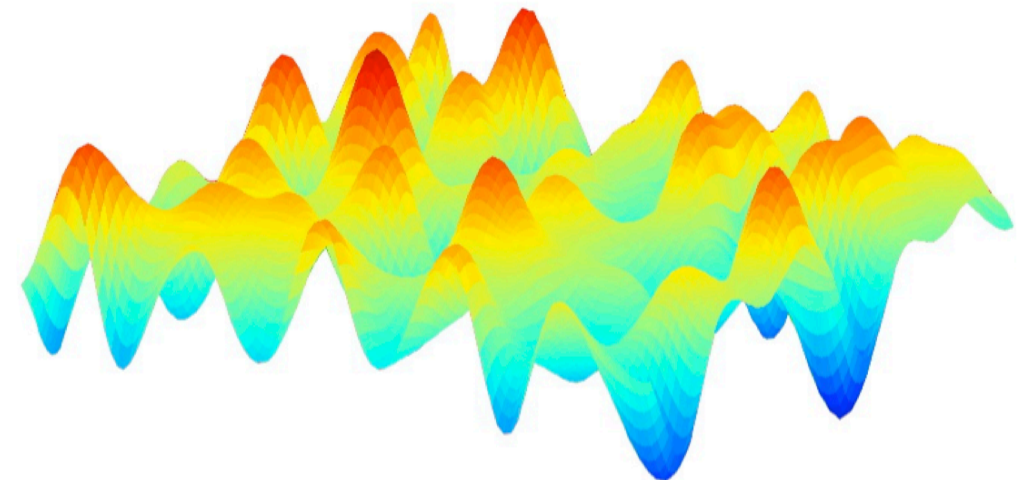
new insight



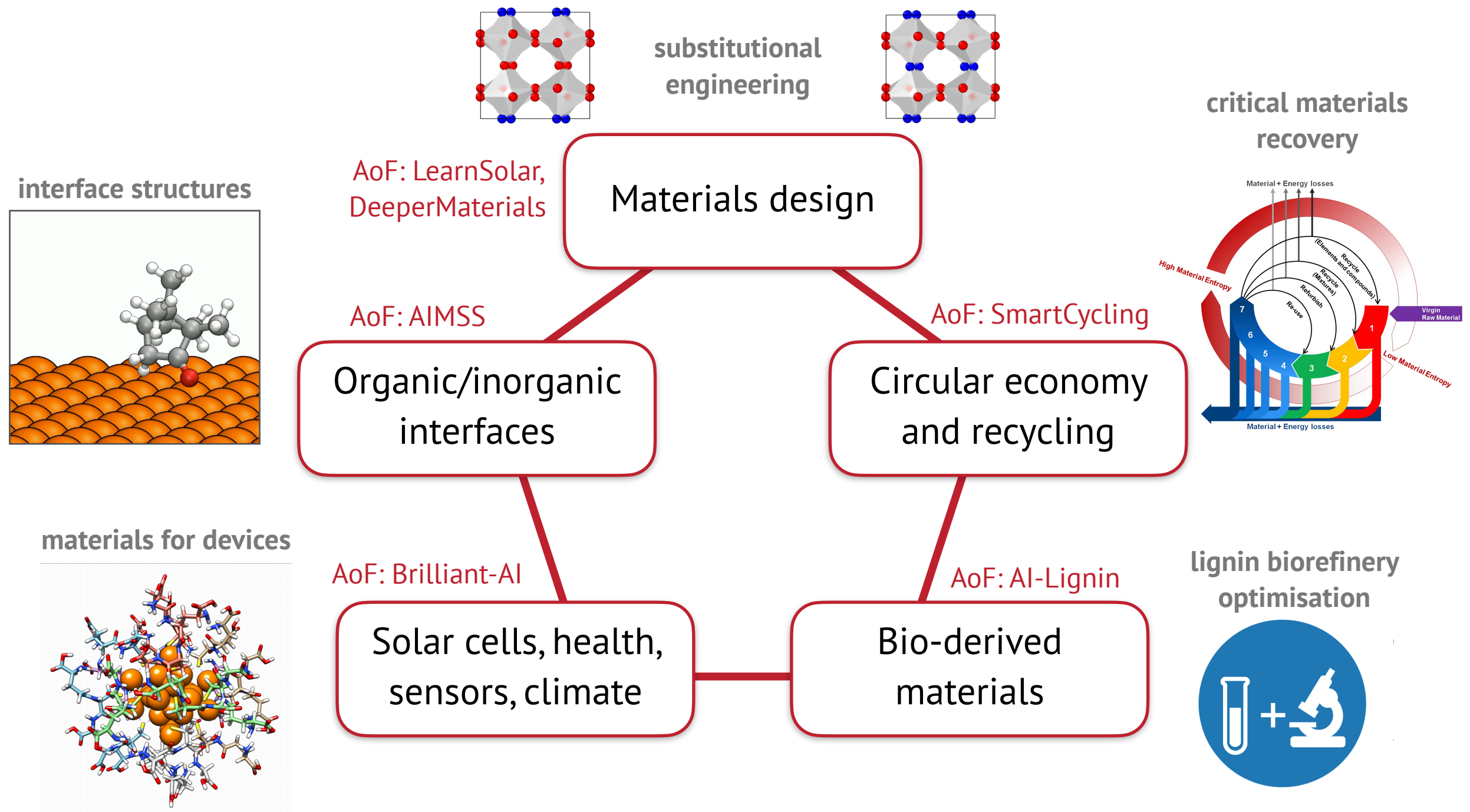
Accelerating discovery in materials science:

- data analytics
- pre-screening
- materials design
- device tuning and optimization
- guiding experimental data collection

processing-structure-property-function
relationships



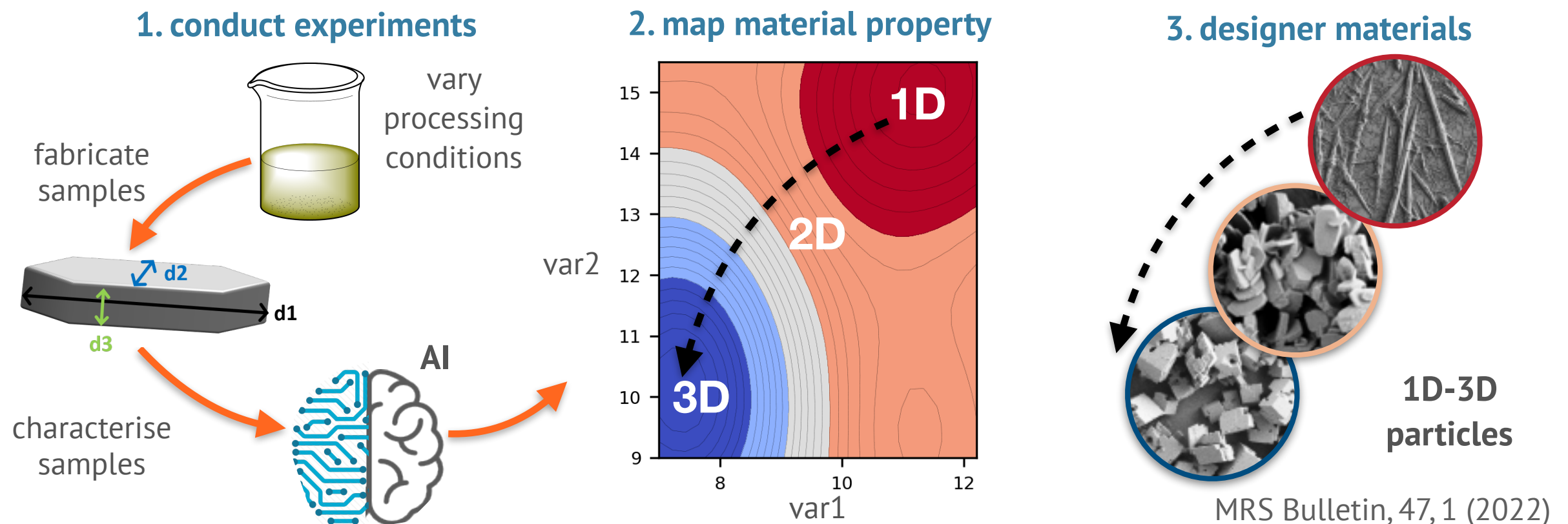
Materials Informatics Laboratory at UTU



Applying AI to refine experiments

Bayesian Optimization **guides experimental data collection** to develop **predictive models** (possibly high-dimensional) and **optimize target properties**, while conducting **as few experiments as possible**.

EXAMPLE: particle morphology engineering (17 samples)



AI-driven optimal outcomes in SmartBIO

Potential applications in SmartBIO

Materials: multi-objective optimisation of bio-derived materials

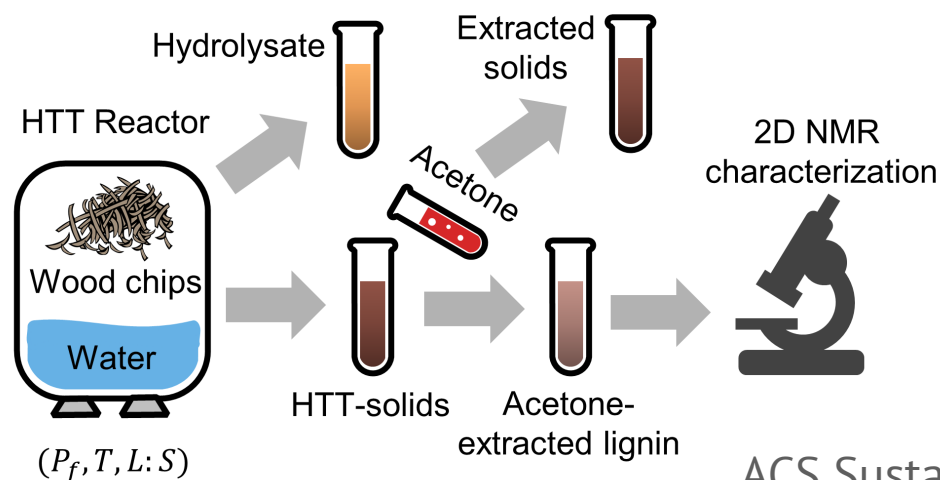
Processes: optimisation of synthesis / functional processes

Devices: device performance optimisation and customisation

EXAMPLE: material process engineering (21 samples)

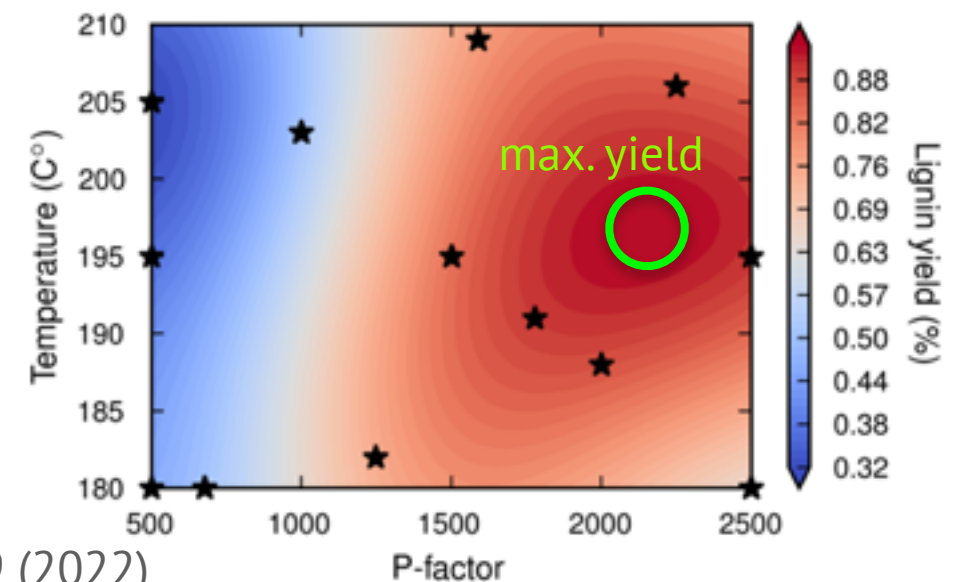
Variables: P-factor, temperature

Target property: Lignin yield



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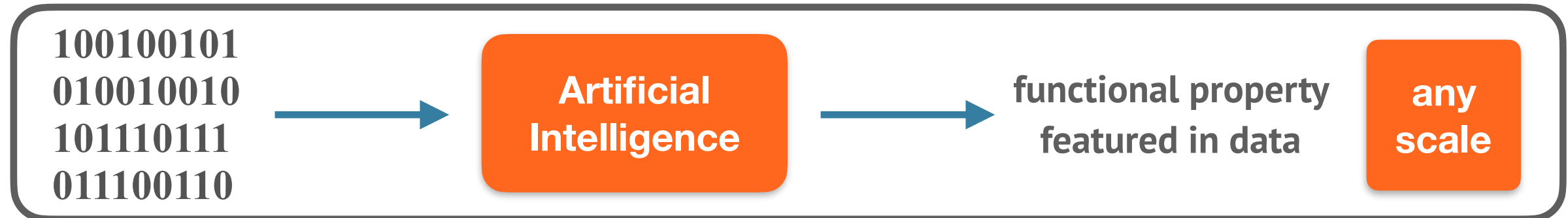
Predictions and optimisation



Why data-driven models?

materials data

data-driven model:



Works with many types of data:

- Material properties (composition, shape, weight...)
- Processing conditions (flow, temperature, pressure...)
- Device parameters (layer thickness, dimensions...)
- Instrument settings

Advantages:

- Sample-efficient and guided data collection (Design of Experiments)
- Yields easily interpretable correlations between target properties and experimental variables.
- Allows multi-objective optimisation