

# Fast optimization of additively manufactured metallic parts with a combination of adaptive feedforward control and numerical simulation (SMARTAM)

## **Academic Partners:**

R. Logé (EPFL), C. Leinenbach & K. Wasmer (Empa), S. Van Petegem (PSI), J. F. Löffler (ETHZ)

## **Industrial Partners:**

PX Group, Heraeus Materials, Patek Philippe, Swatch Group

## **Potential other industrial partners:**

Richemont-Varinor, Rolex SA

## Project overview - Motivation

Laser Powder Bed Fusion (LPBF) is a layer by layer deposition additive manufacturing technique

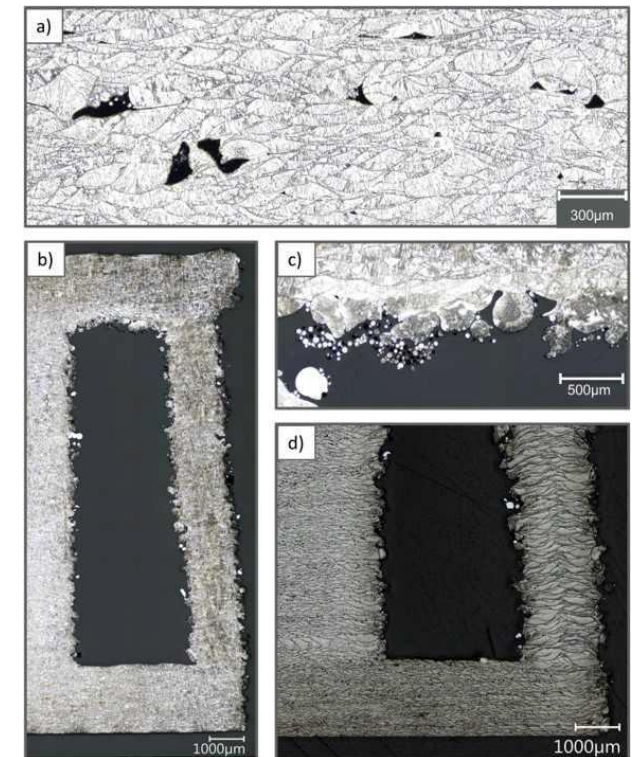
Single sets of process parameters are defined for a given part despite its intricate geometry.

Dependency of heat flux on geometrical features results in variations in melt pool geometry which may lead to the formation of undesirable defects.

Variations in the imposed local thermal history additionally can result in undesirable microstructure.

### REMEDY

**Part specific and location specific** process parameters can be derived from numerical simulations and machine learning algorithms.

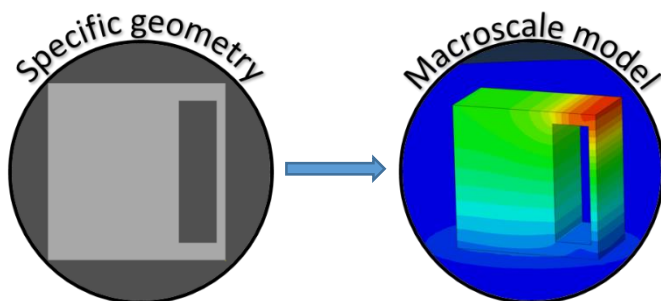


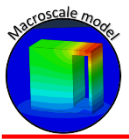
C.L. Druzgalski, et al. Additive Manufacturing, (2020)

## Project work flow

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- Core workflow
- Verification/improvement
- Training – A priori

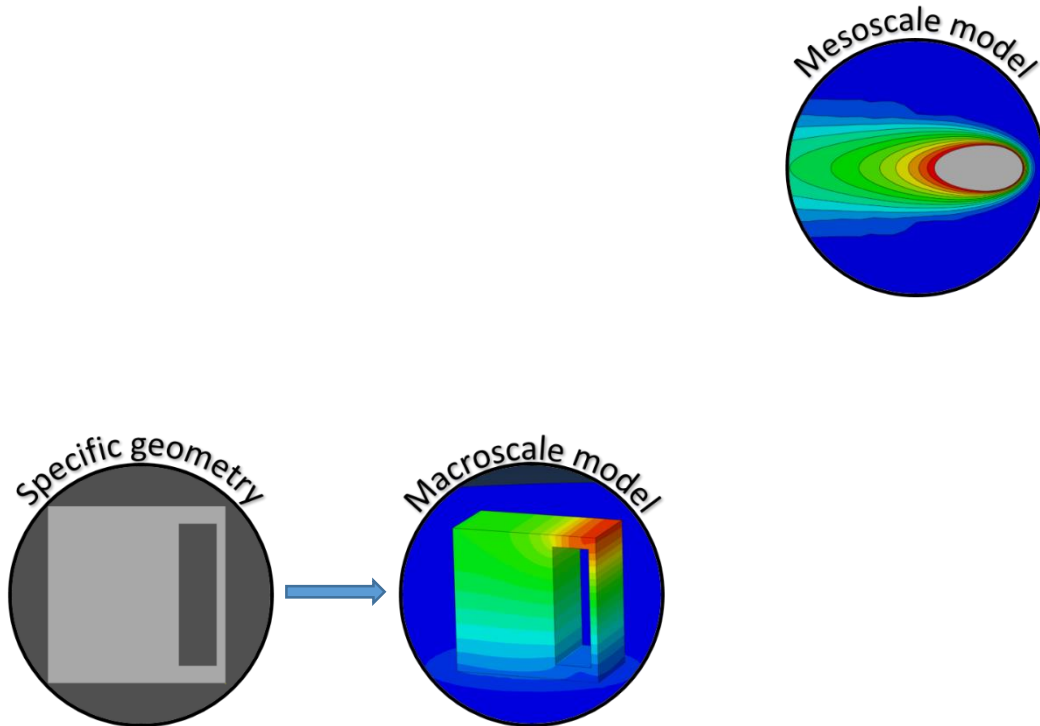


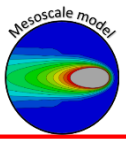


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Simulation in **melt pool scale**  
to obtain melt pool geometry

Design of Experiment (DOE):

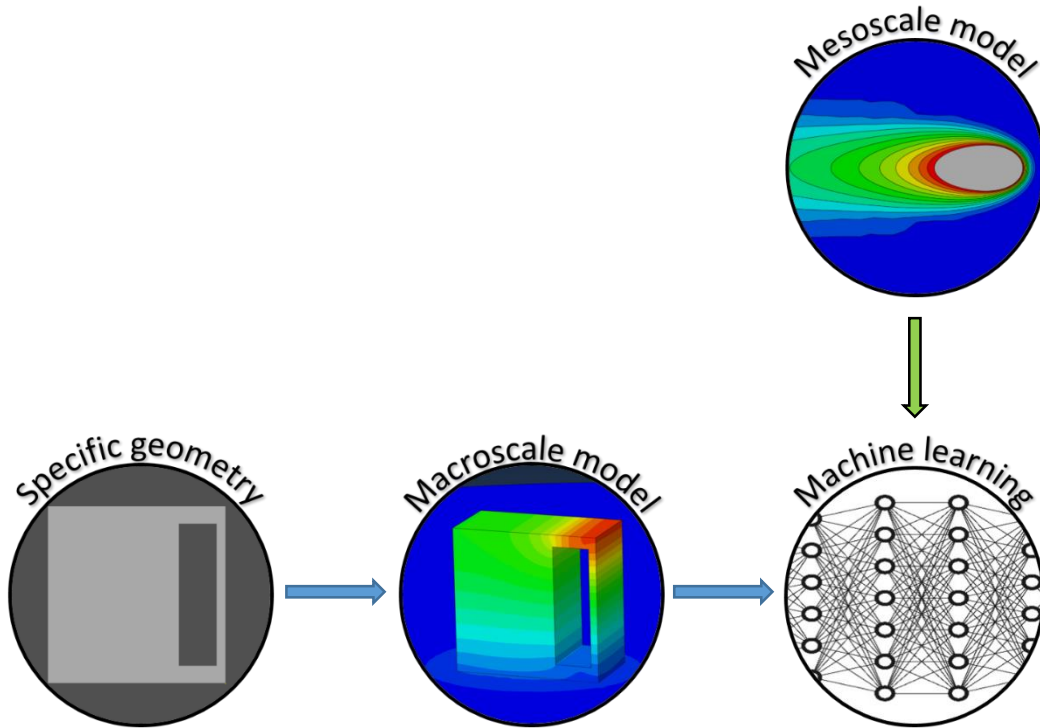
- $T_0$ ; initial temperature
- $P$ ; Laser power
- $v$ ; Scanning velocity



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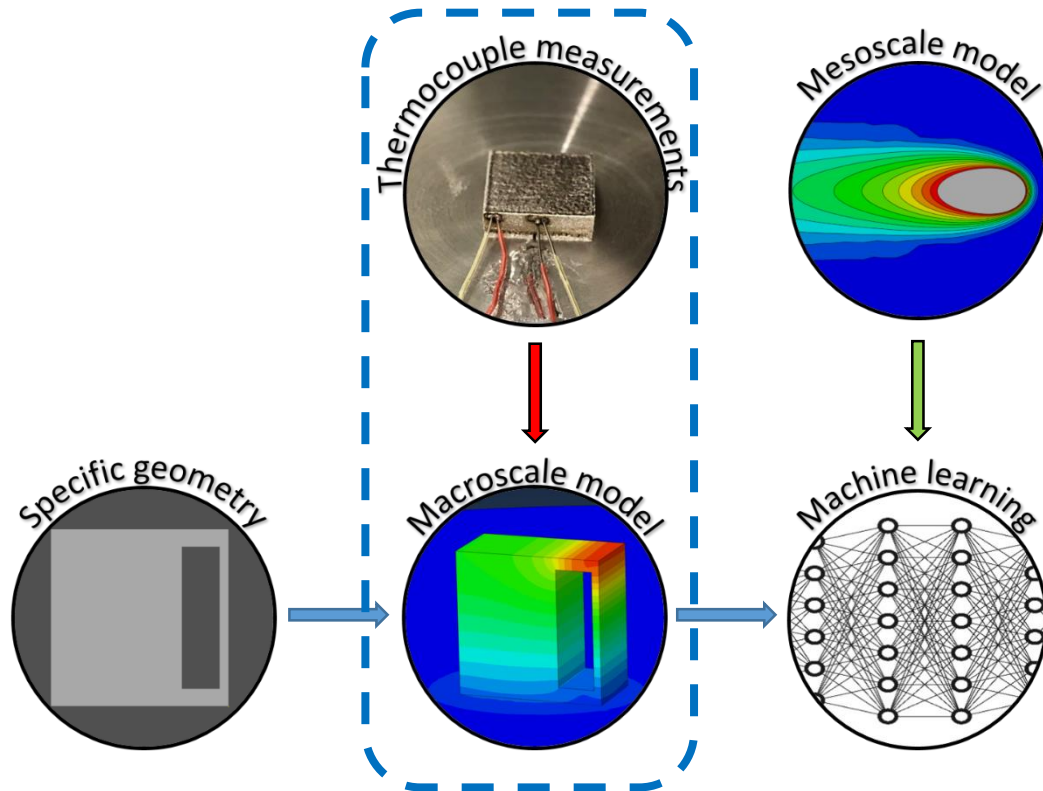
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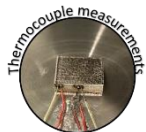


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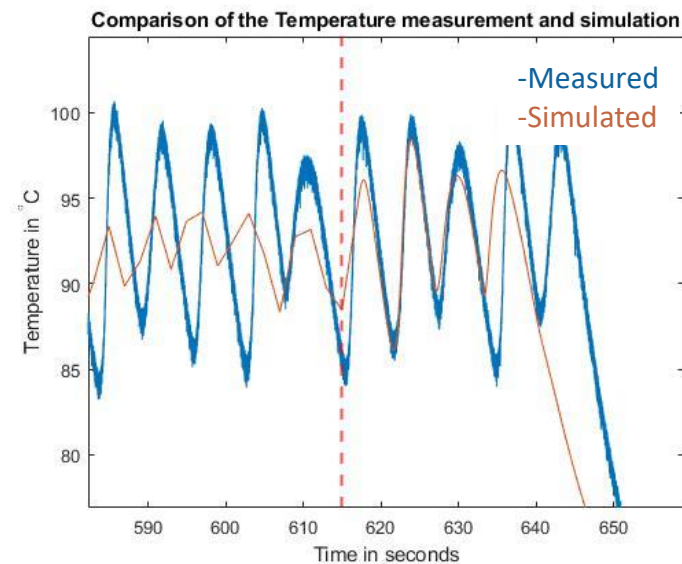
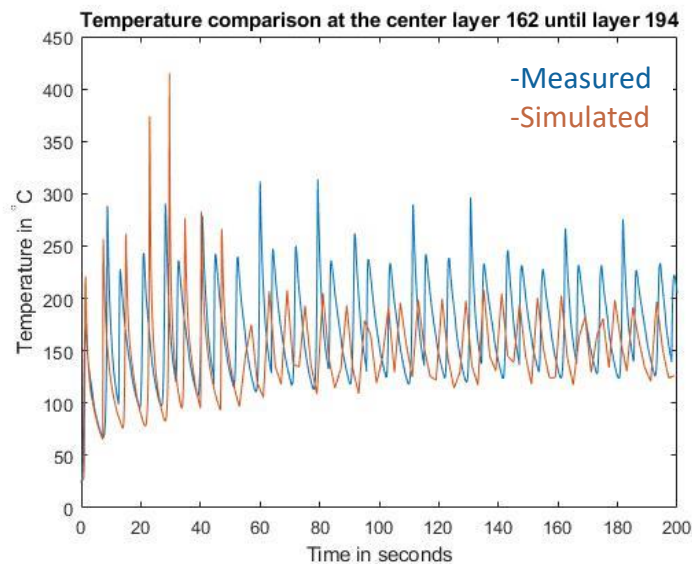
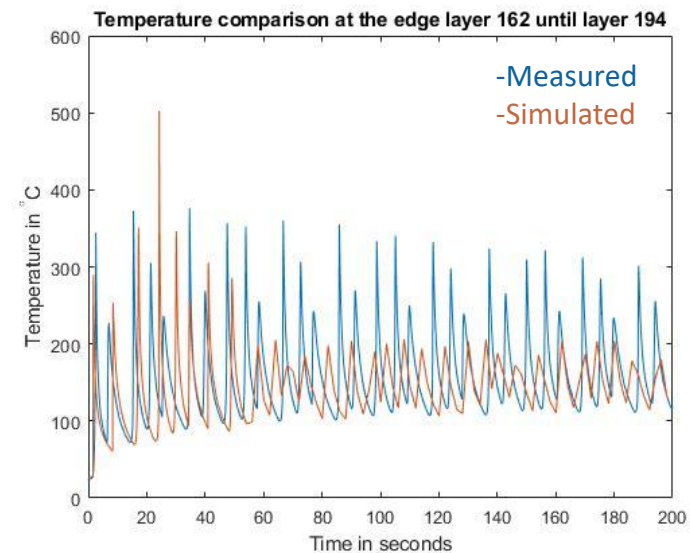
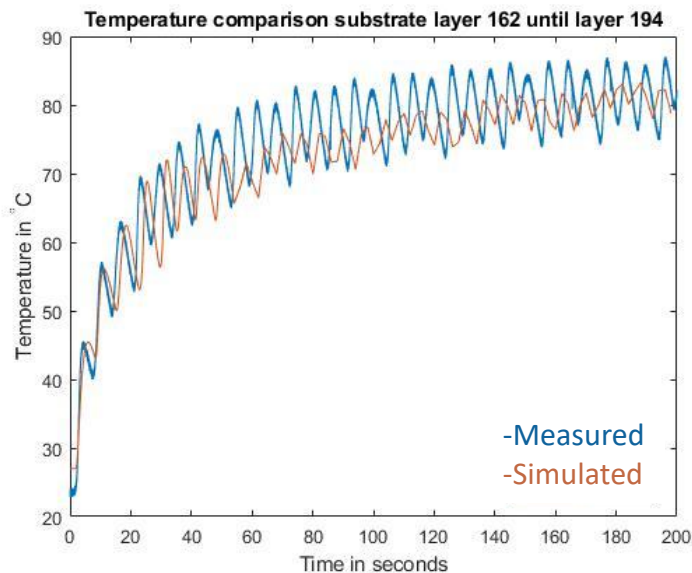
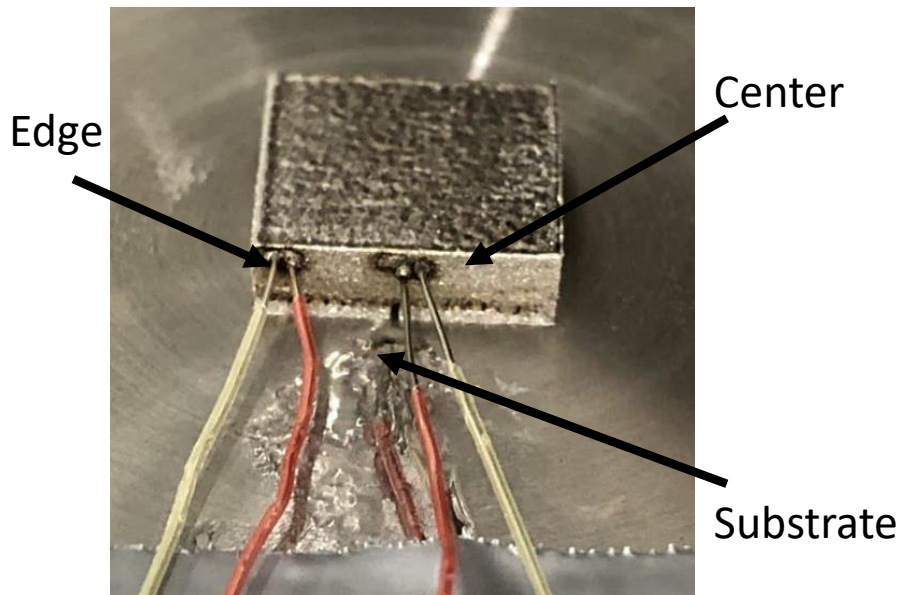






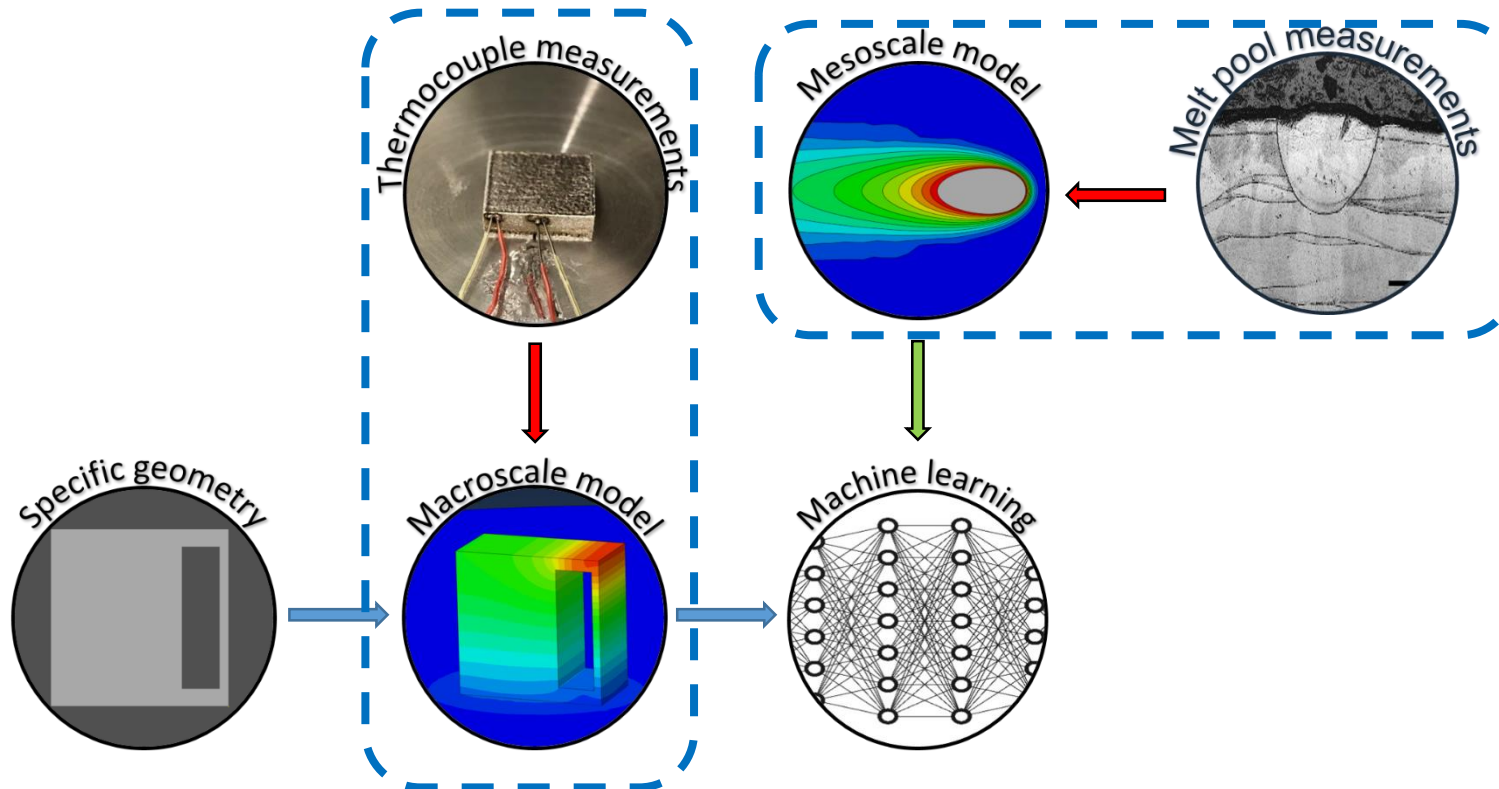
# Macro-scale simulation - Calibration

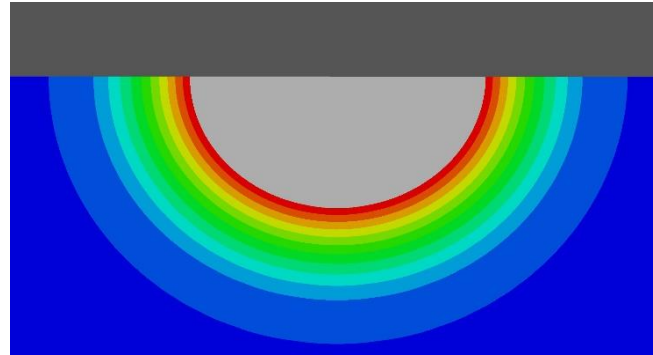
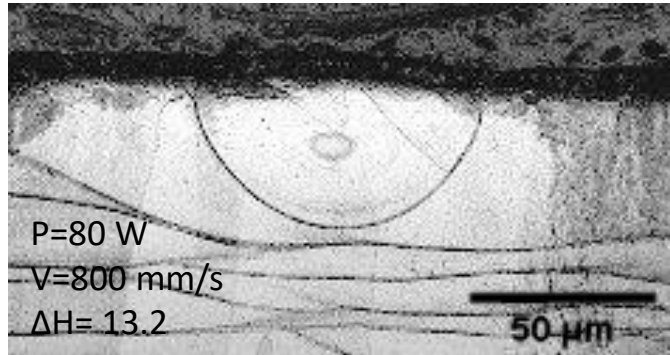
Credits: Lucas Schlenger, Jamasp Jhabvala



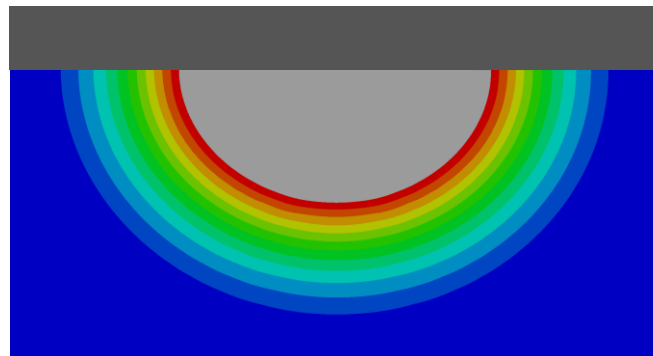
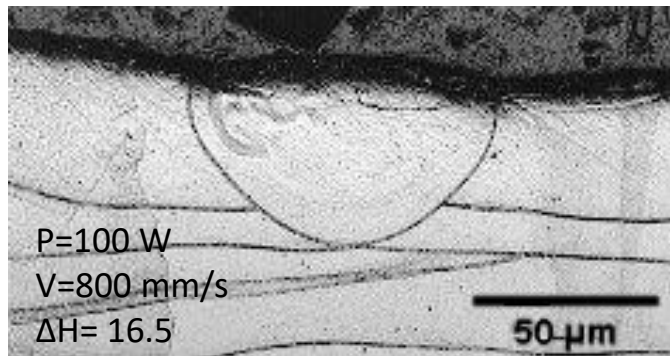
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	Width (μm)	Depth (μm)
Measured	69.1	36.3
Simulated	71.6	33.9



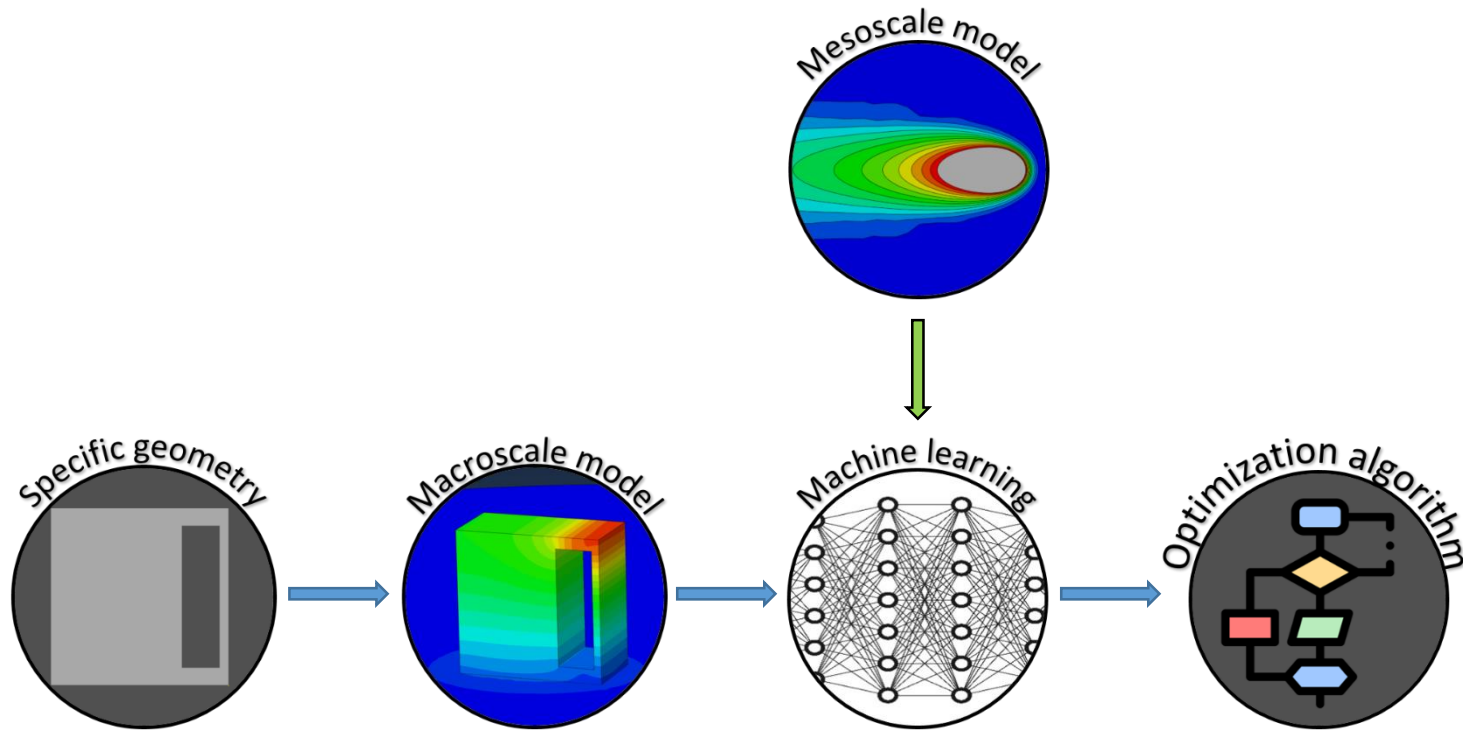
	Width (μm)	Depth (μm)
Measured	79.3	46.5
Simulated	80.4	37.1

Comparison study for 10 different samples

	Width (μm)	Depth (μm)
MAD	3.8	7.8
Bias	-0.02	-6.5

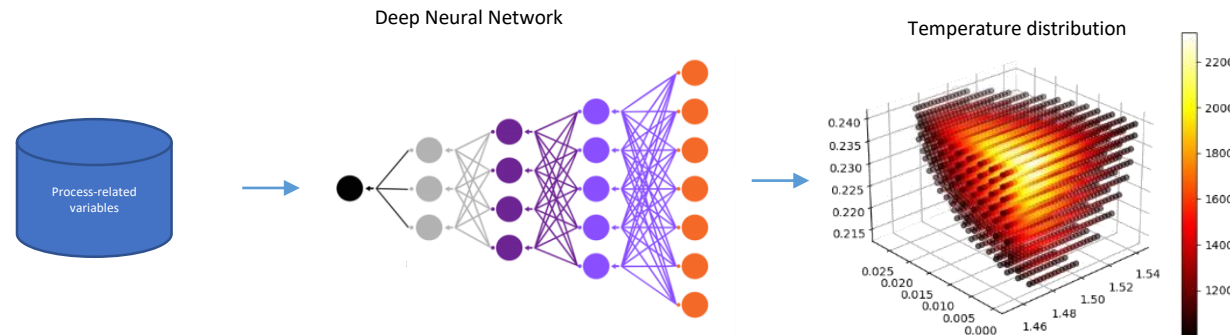
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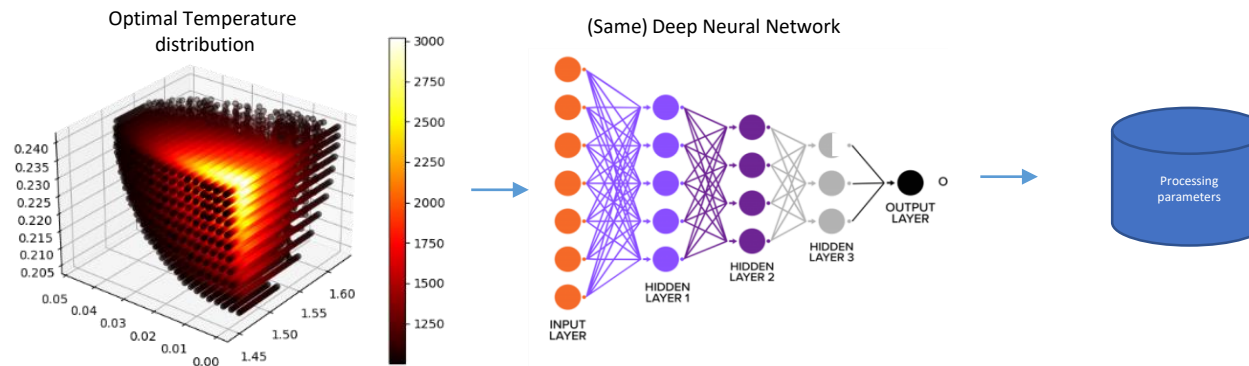


The trained Deep Neural network can solve two problems

1. The Forward problem — replacing the steady state model:

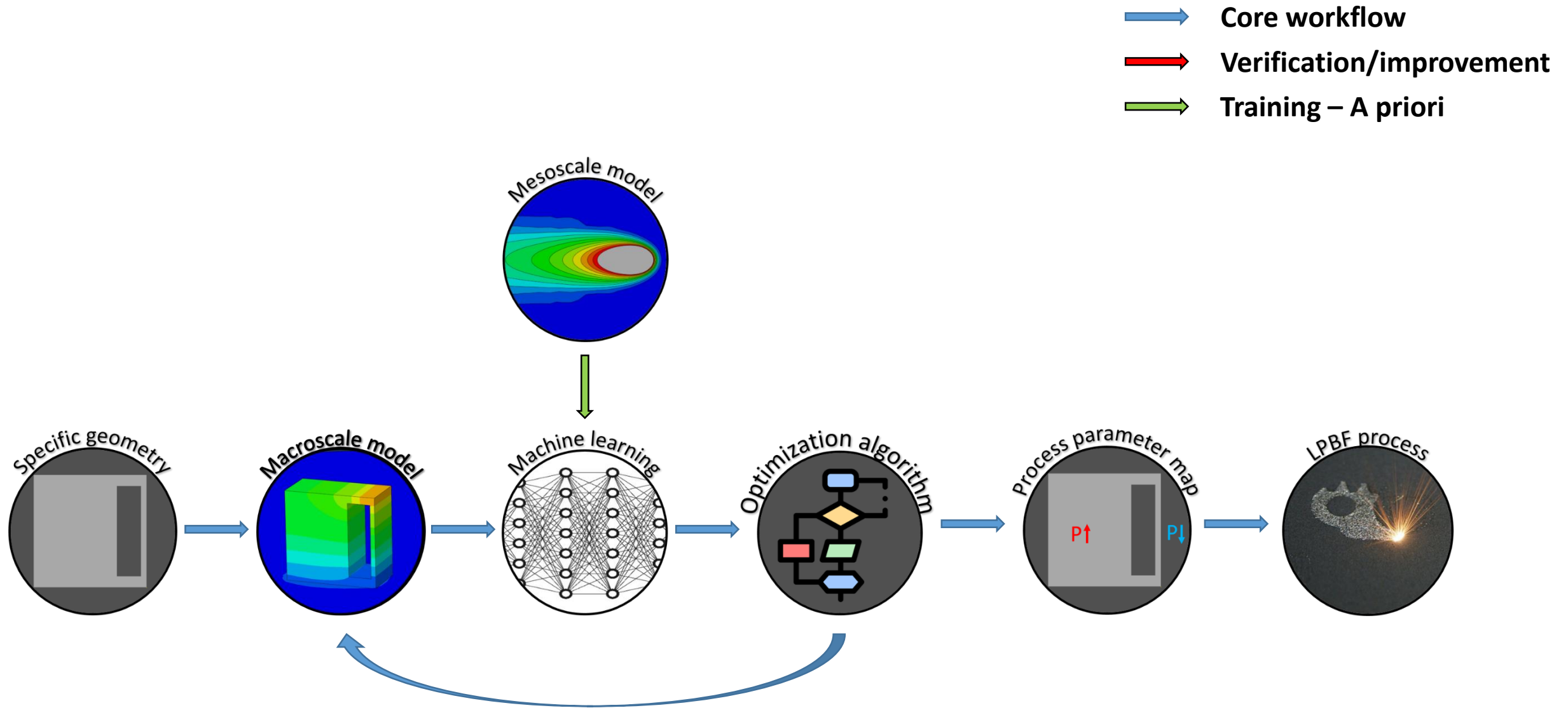


2. The Inverse problem — performing the desired optimization:

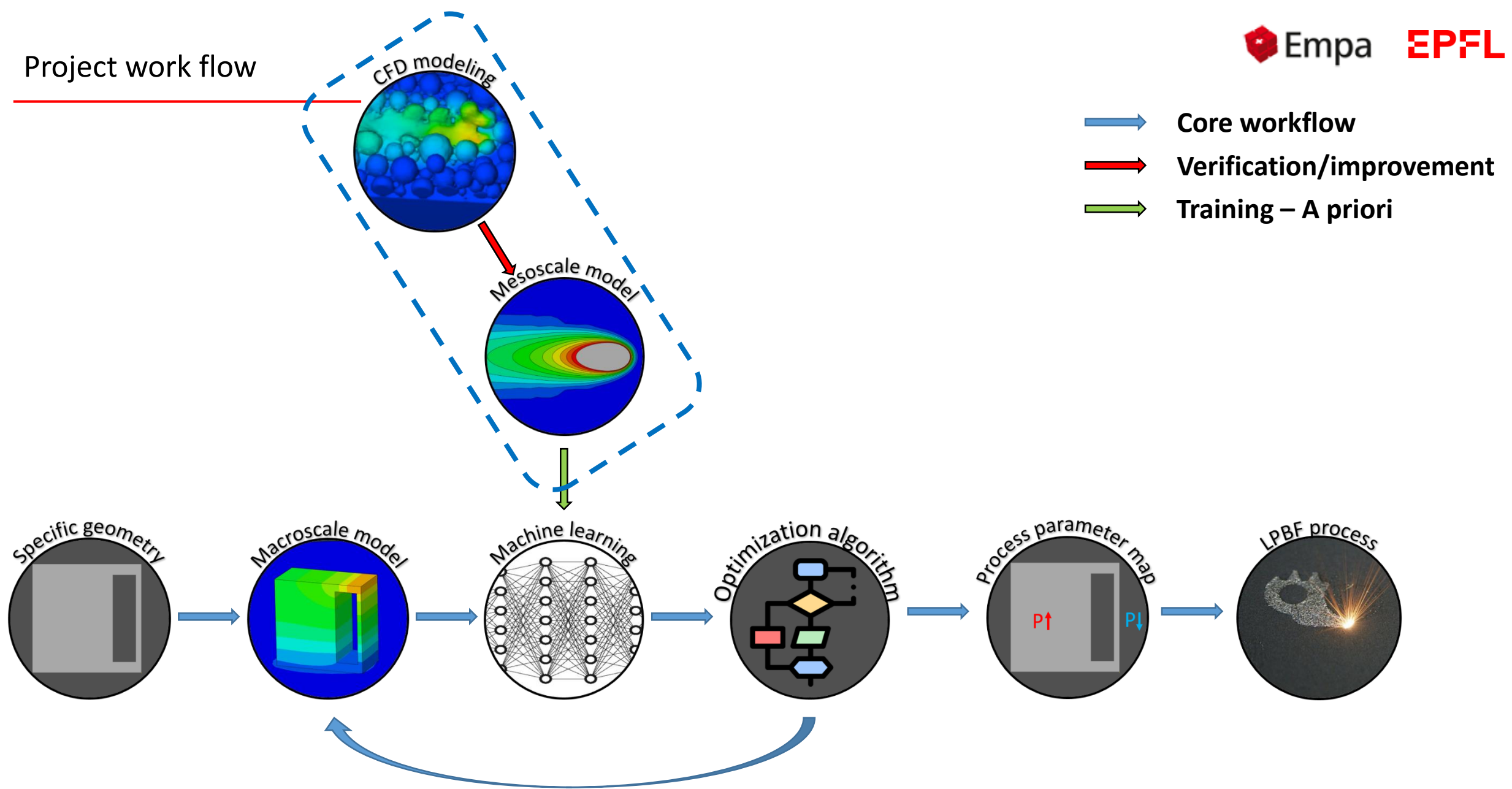


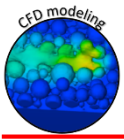


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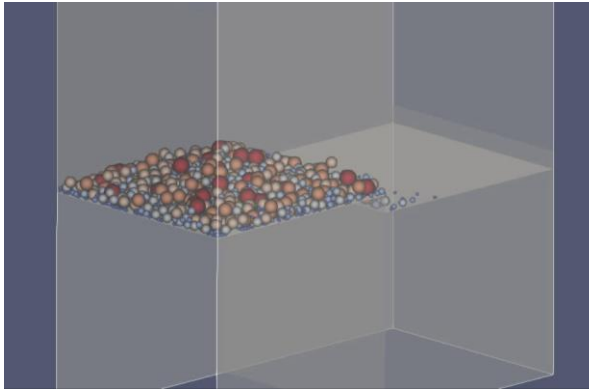
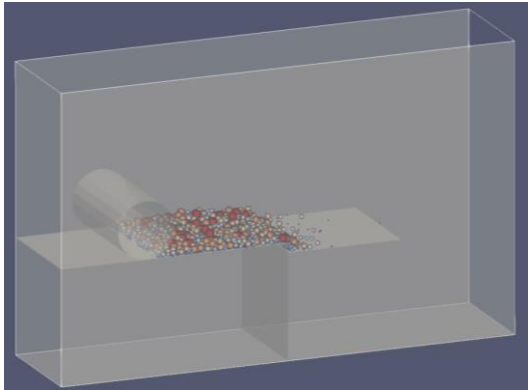
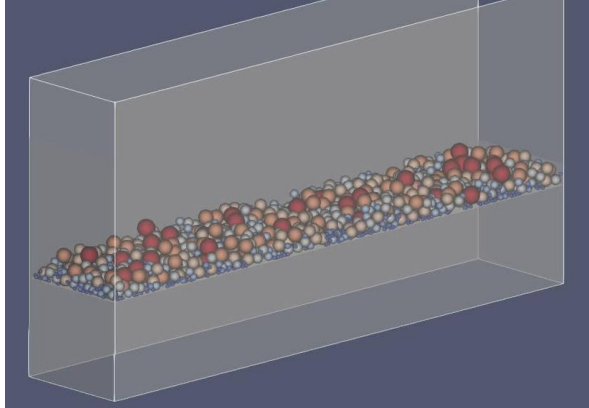
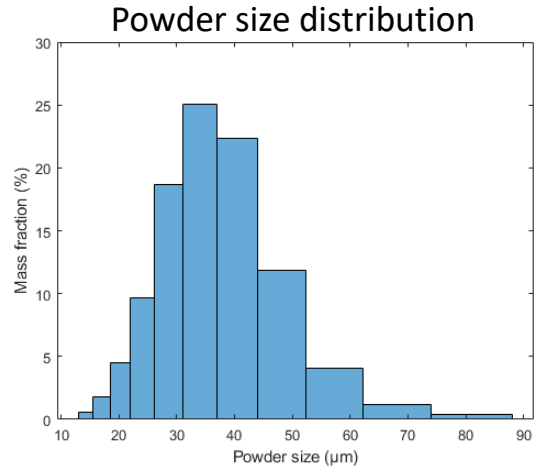
# Project work flow



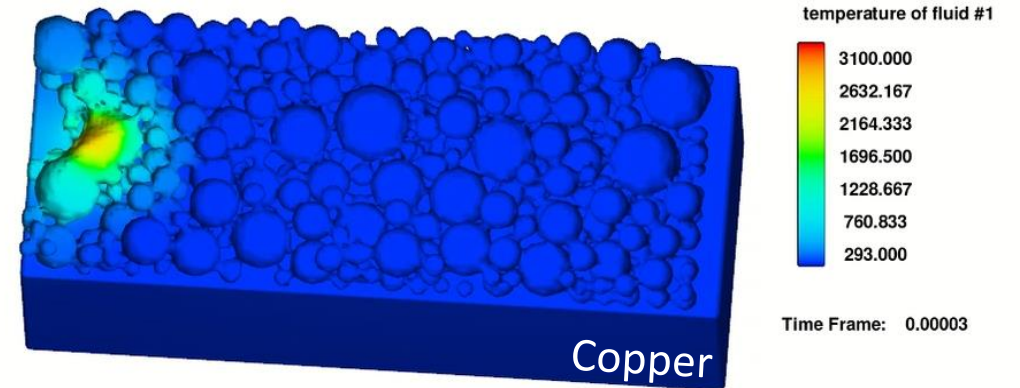
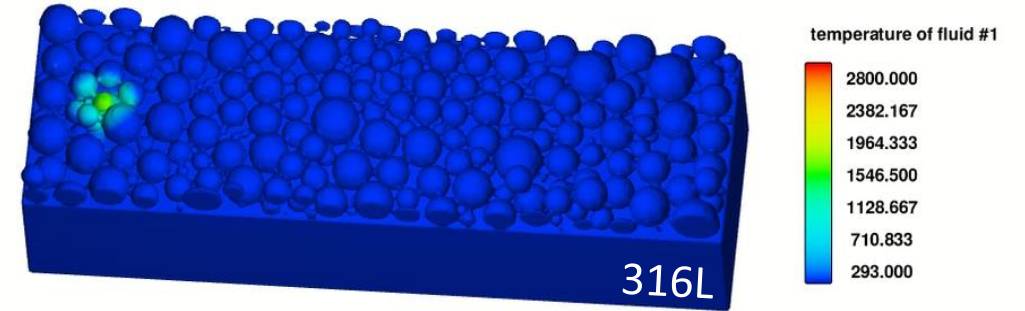


# Computational Fluid Dynamics (CFD) Modeling

Credits: Jian Yang, Christian Leinenbach



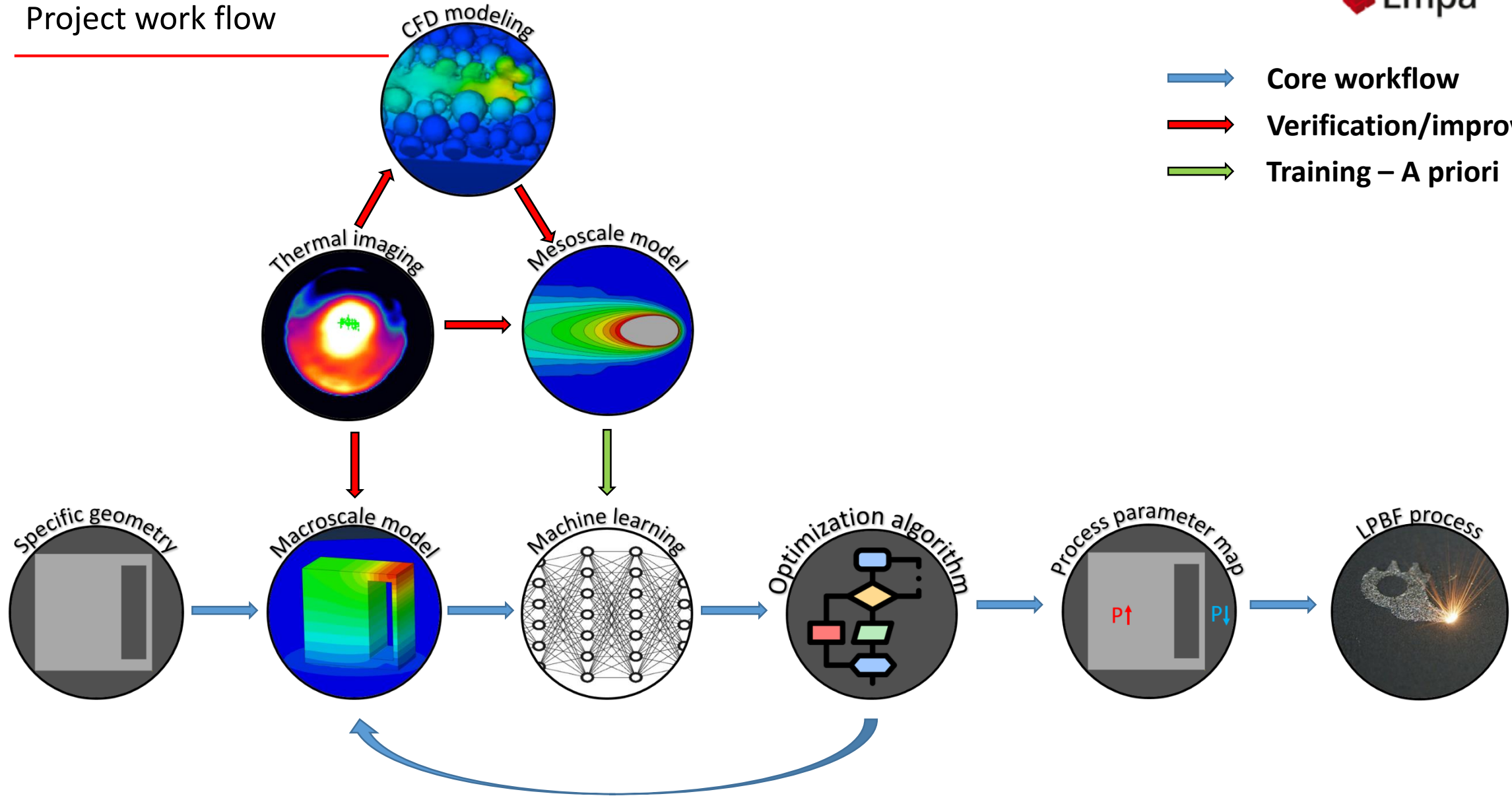
Different re-coating simulations (DED)

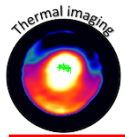


CFD simulations

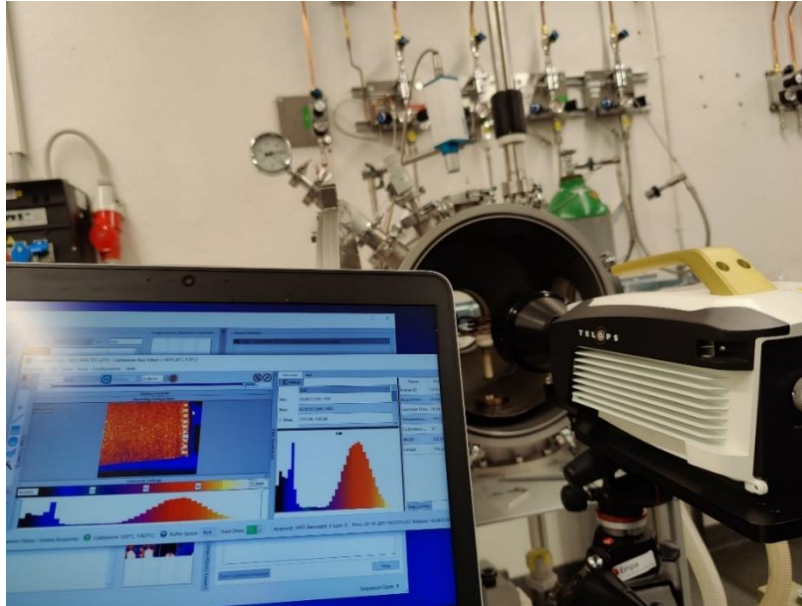


# Project work flow



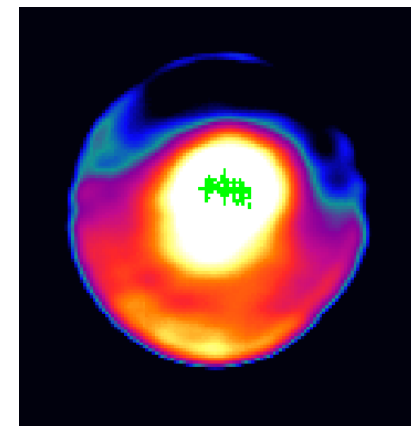
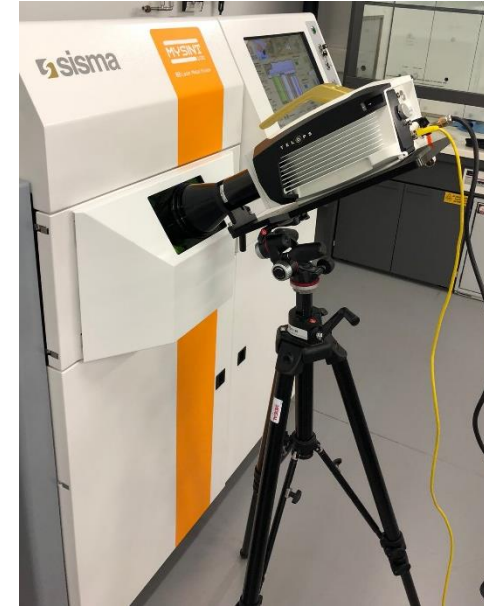


## Experimental setup



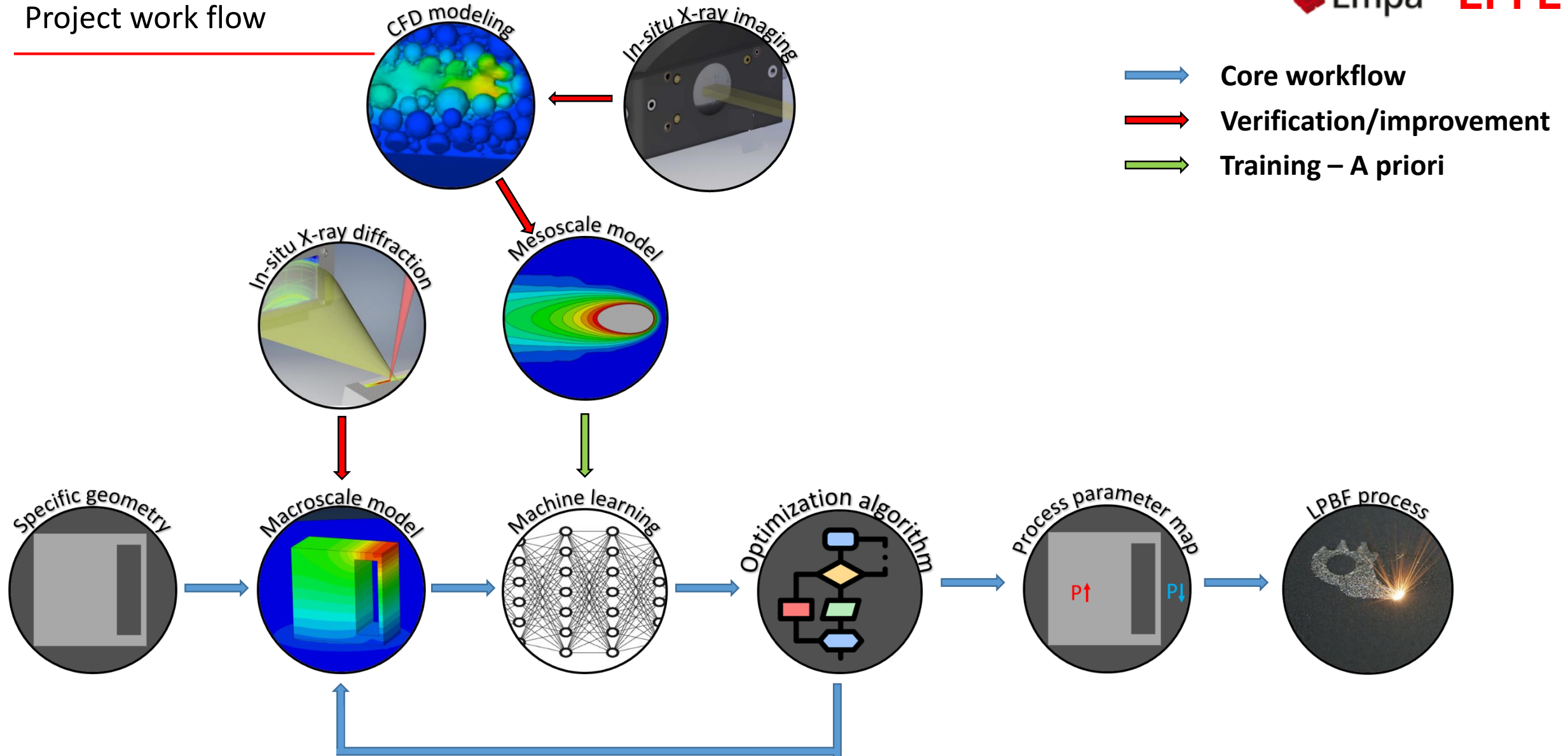
Calibration of emissivity  $\varepsilon$  with  
induction furnace and pyrometer

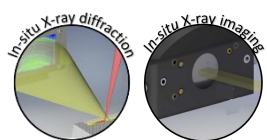
Close collaboration between EPFL and Empa



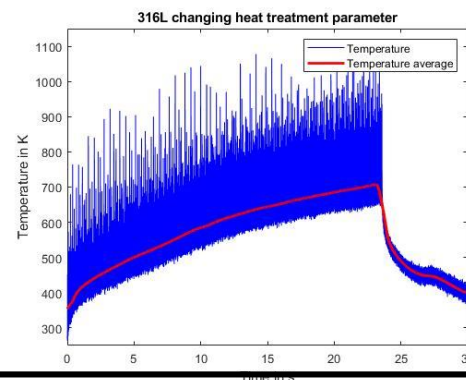
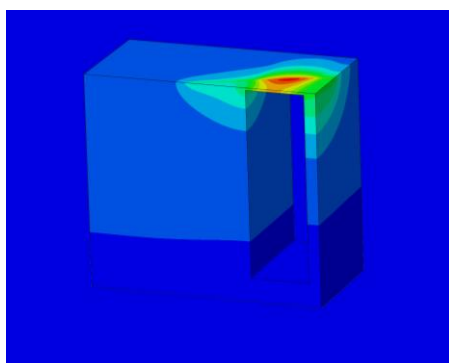
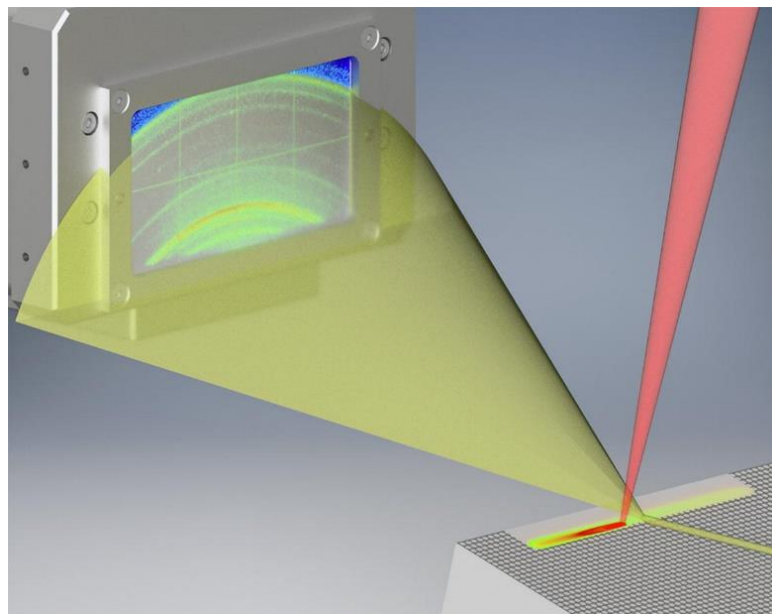
Melt pool measured  
on experimental LPBF  
machine

# Project work flow

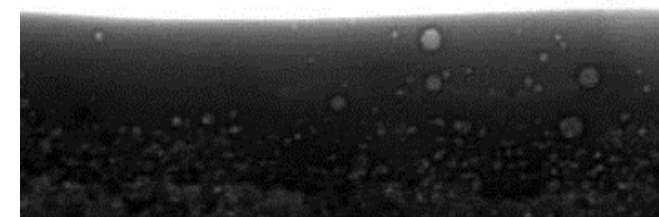
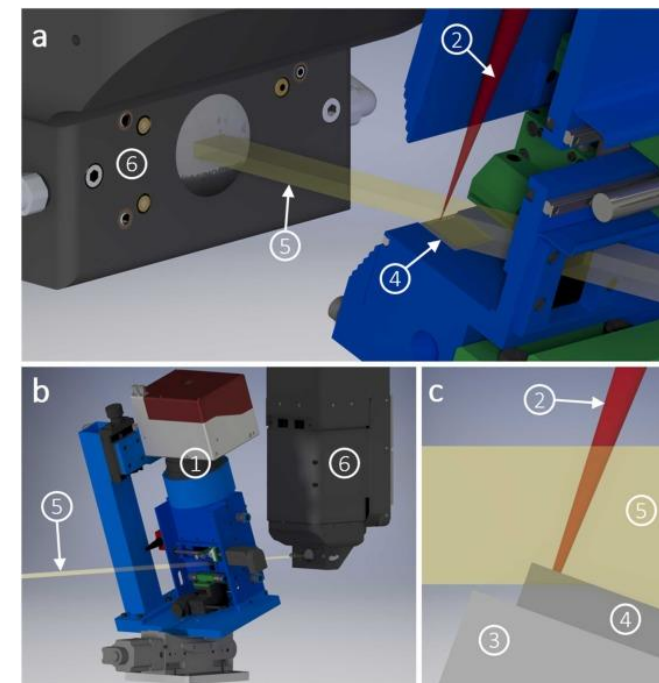




# SLS Beamtime proposals

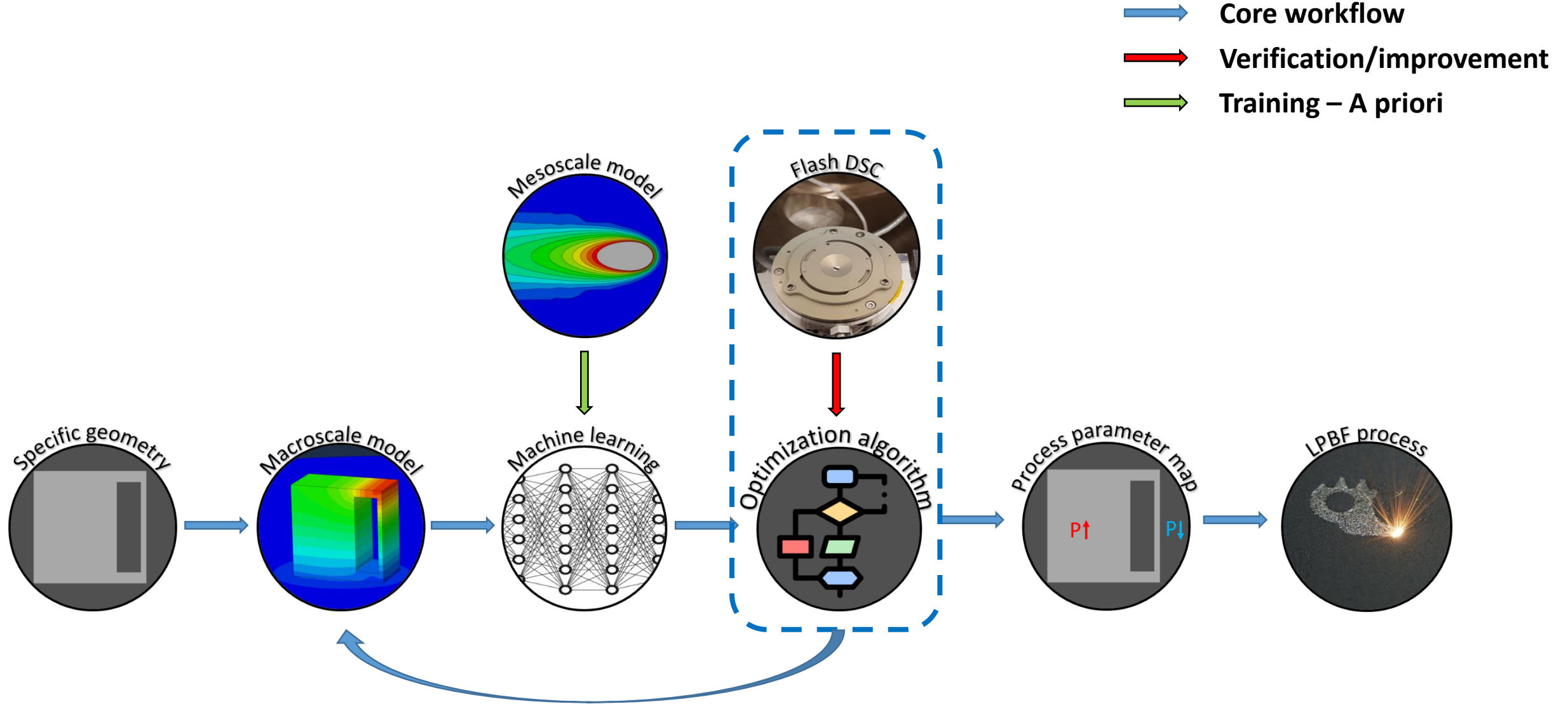


*Operando X-ray diffraction*



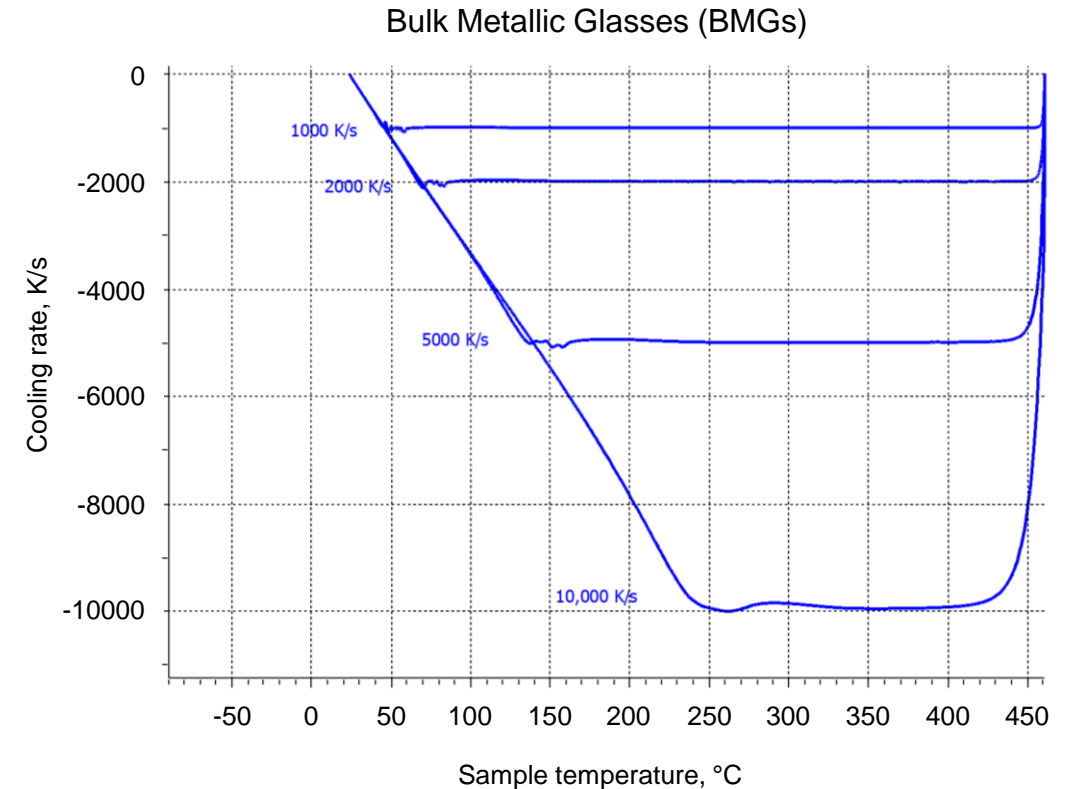
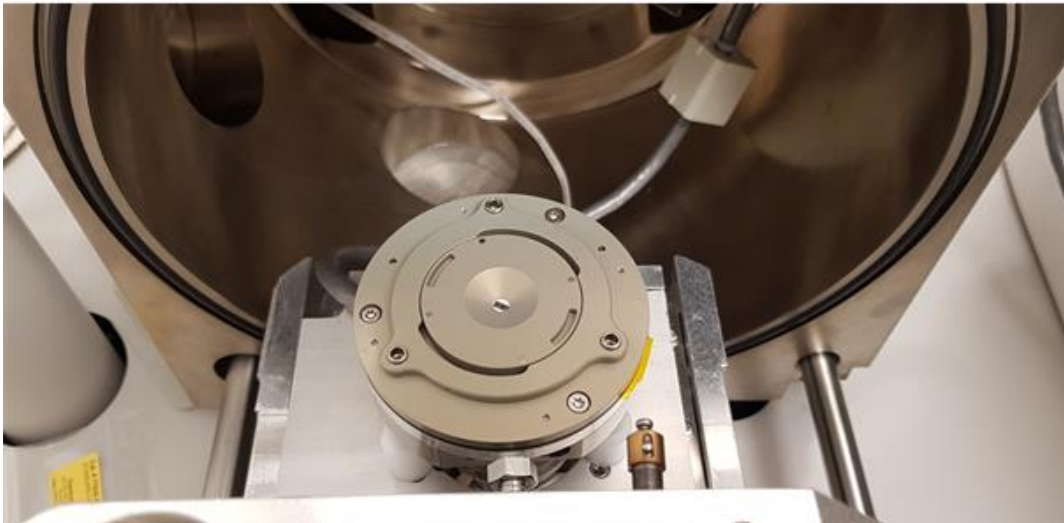
*Operando X-ray imaging*







- calorimetry is sensitive, but not specific to structural changes  
→ new *in situ* FDSC within a SEM (LMPT + ScopeM – ETH Zürich)
- Thermo Fisher Scientific ESEM Quattro and Flash DSC 1 with UFS1 sensors (Mettler-Toledo)
- SE detector, high frame rate
- BSE, EDX planned in future
- Limiting factor – slow signal collection in SEM



FDSC-SEM – cooling conditions



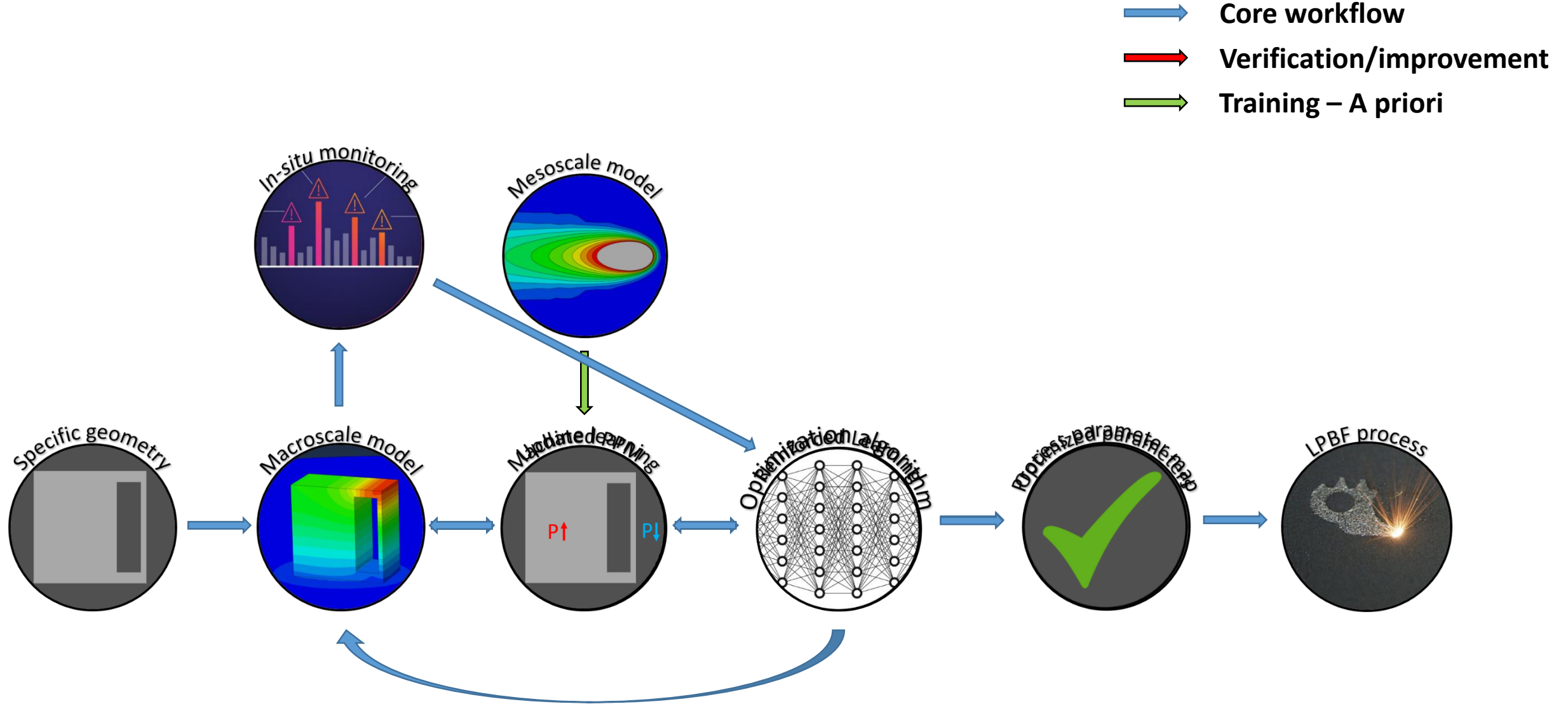
- Structural changes AND simultaneous acquisition of heat flow signal
- Heating curves at 5 K/s after cooling from melt at various cooling rates were recorded
- Microstructural changes not seen in FDSC curves and thermal events observed without microstructure change

→ **complementary methods**



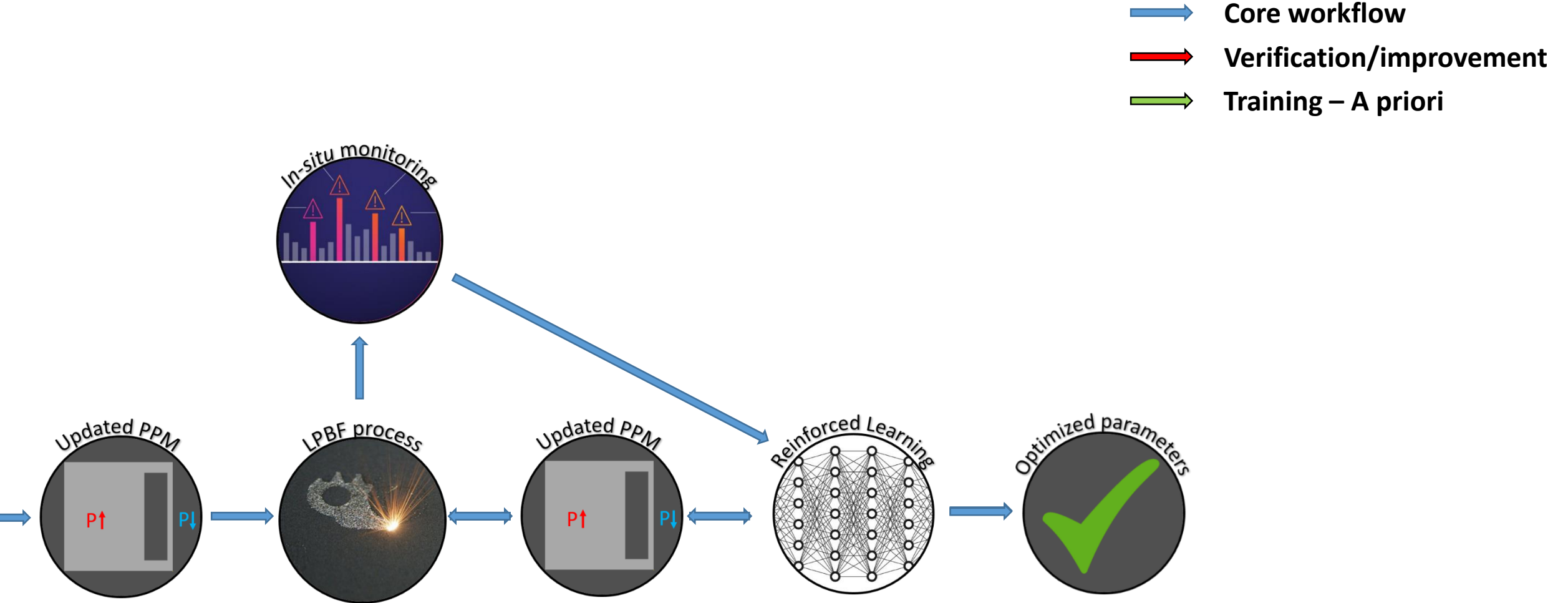
Eutectic AuGe alloy previously cooled at 10 000 K/s (ballistic) – heating rate 5 K/s

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