

Real-time Digital Twin Development for Additive Manufacturing

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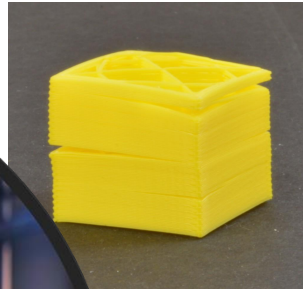
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1. Motivation & Problem Statement

Bonding



Crack



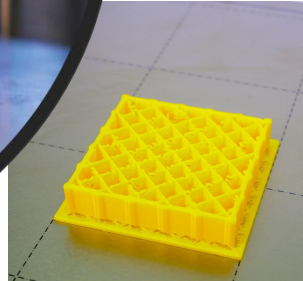
Warpage



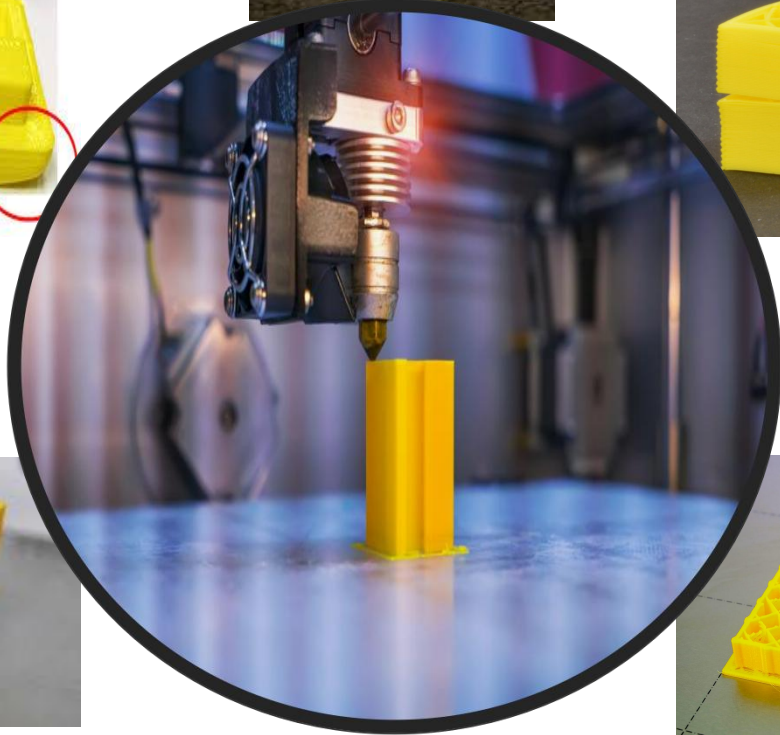
Geometry



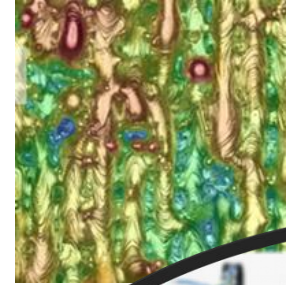
Infill



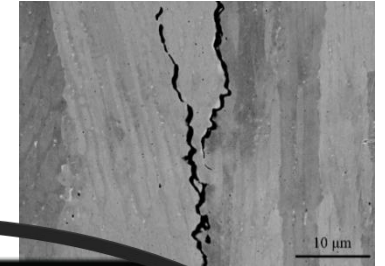
FFF



Surface Roughness



Crack



LPBF

Unfused Powder



Porosity



1. Motivation & Problem Statement



tensile test



bending test



compression test

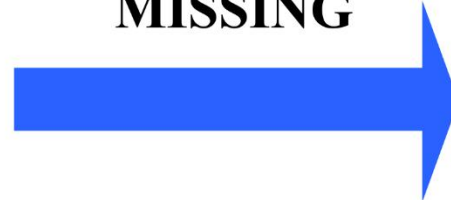
1. Motivation & Problem Statement

Gap in Current Additive Manufacturing Monitoring

Existing monitoring systems

- Mostly post-process inspections
- Limited real-time defect detection
- Poor integration of structural validation

WHAT'S MISSING



Integration of sensing, ML, and validation for real-time quality assurance



My intelligent AM monitoring & validation system

- Multi-sensor fusion
- Machine learning-driven detection
- Real-time structural validation
- Decision-making-in-the-loop



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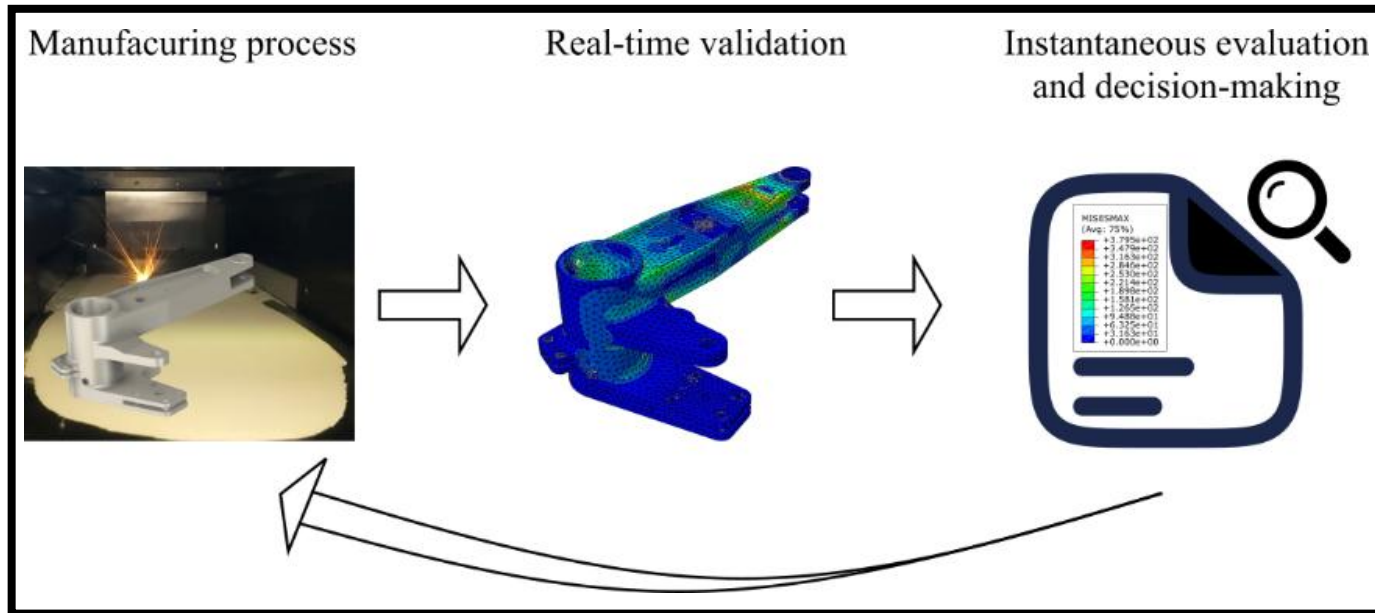
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2. Research Vision & Objective

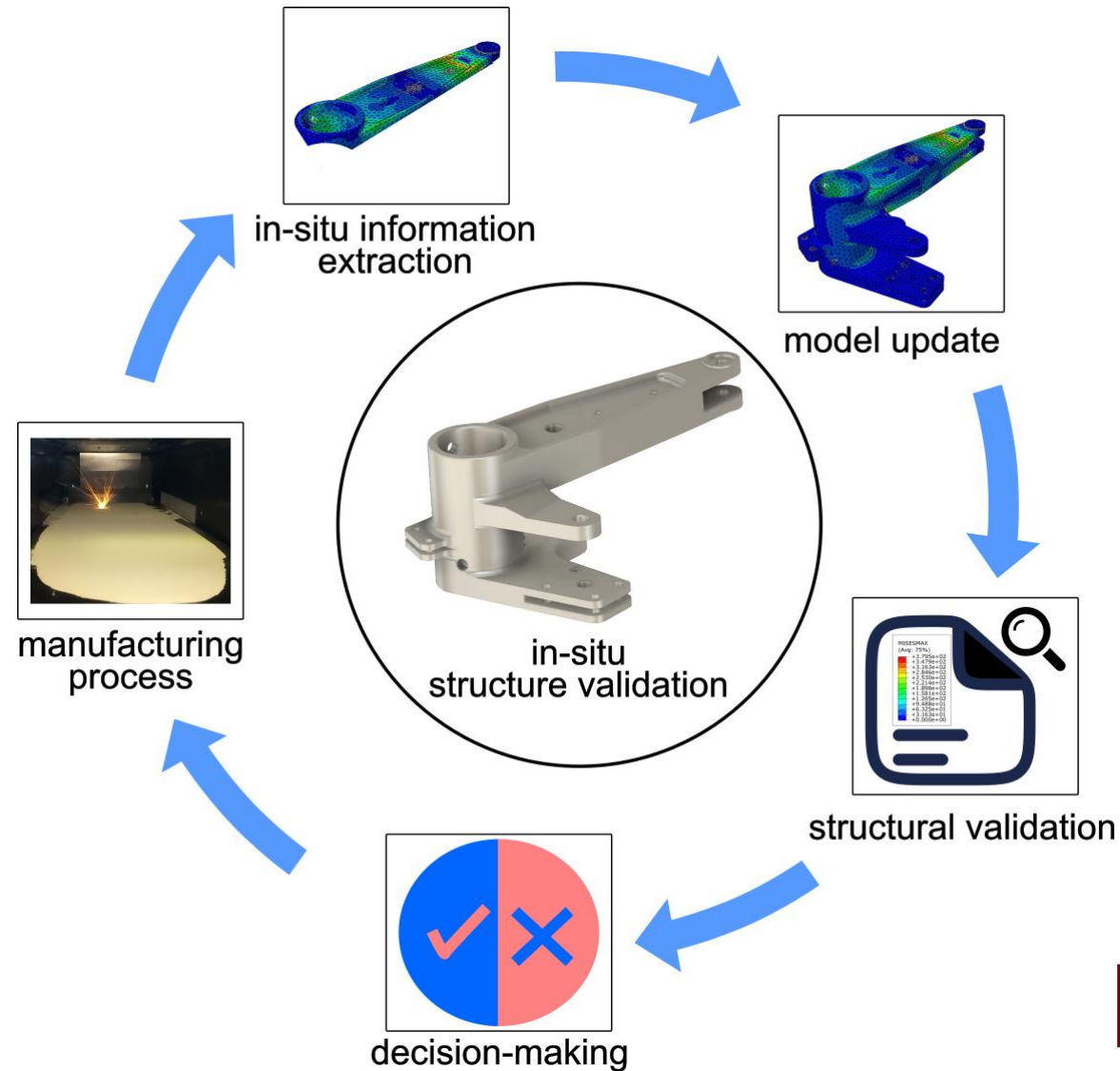


pull off and ready for use



**Molinaroli College of
Engineering and Computing**
UNIVERSITY OF SOUTH CAROLINA

2. Research Vision & Objective



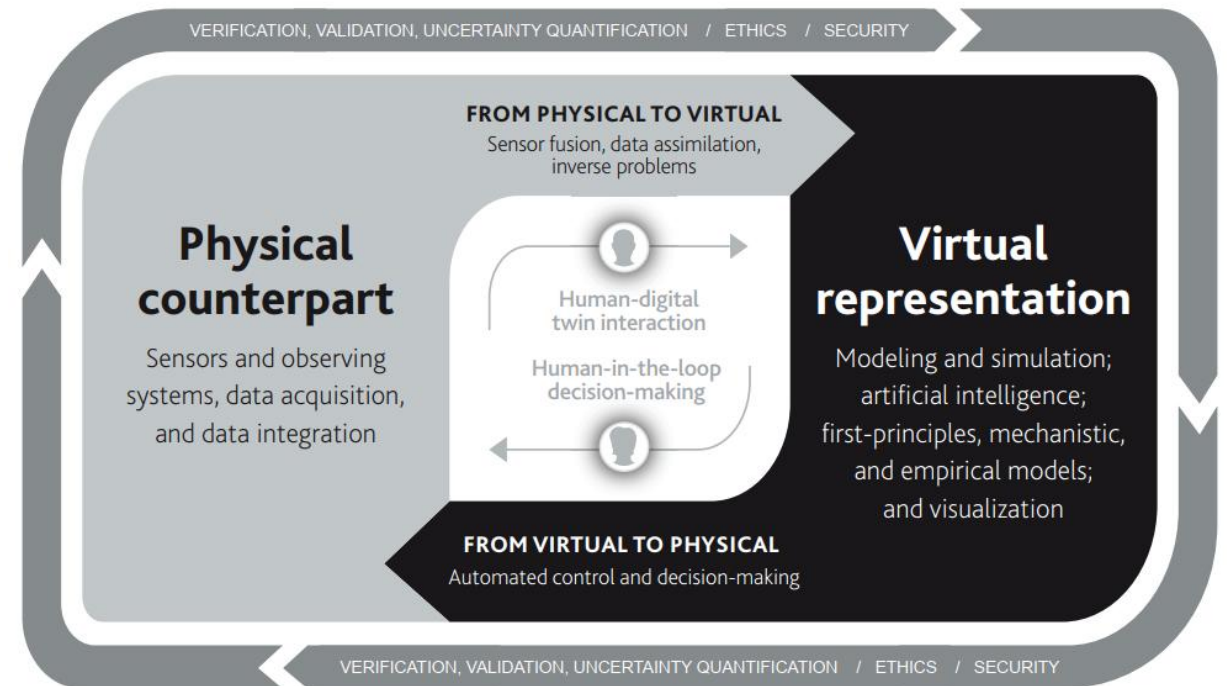
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3. Digital Twin

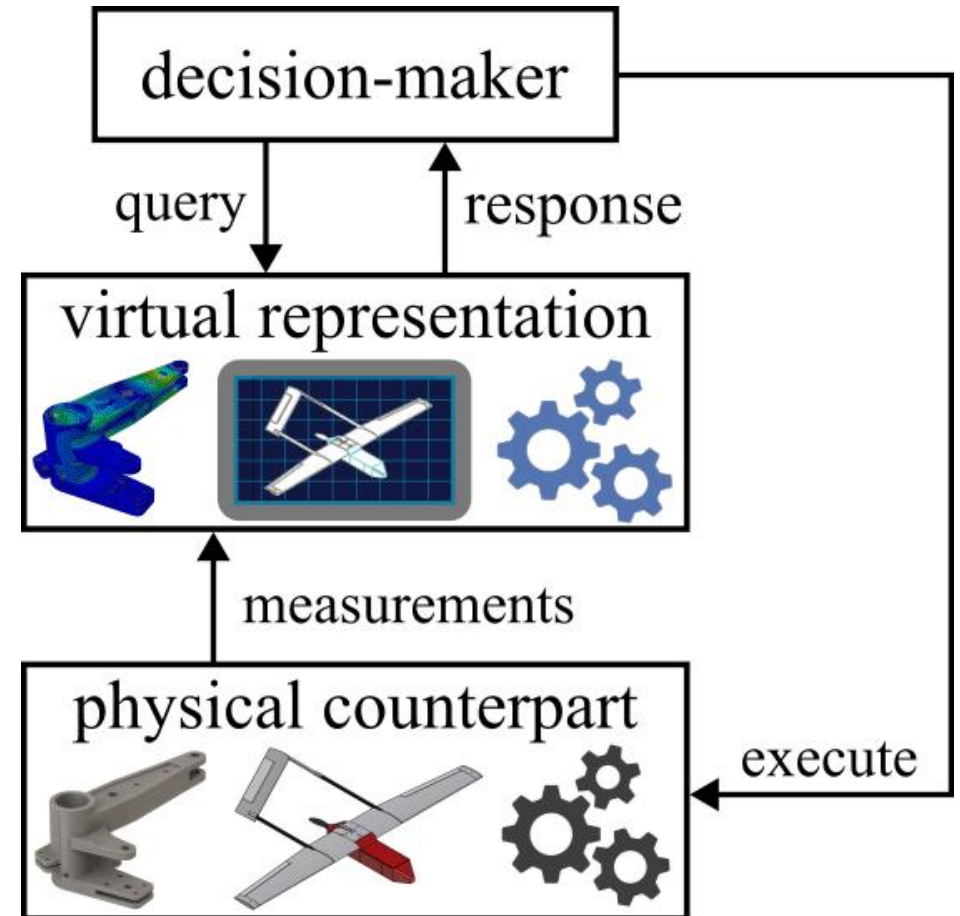
What a digital twin is: A live computer model of a real system that stays synced with sensor data. Not just CAD.

- Information flows bidirectionally between the virtual representation and physical counterpart.
- These information flows may be through automated processes, human-driven processes, or a combination of the two.



3. Digital Twin

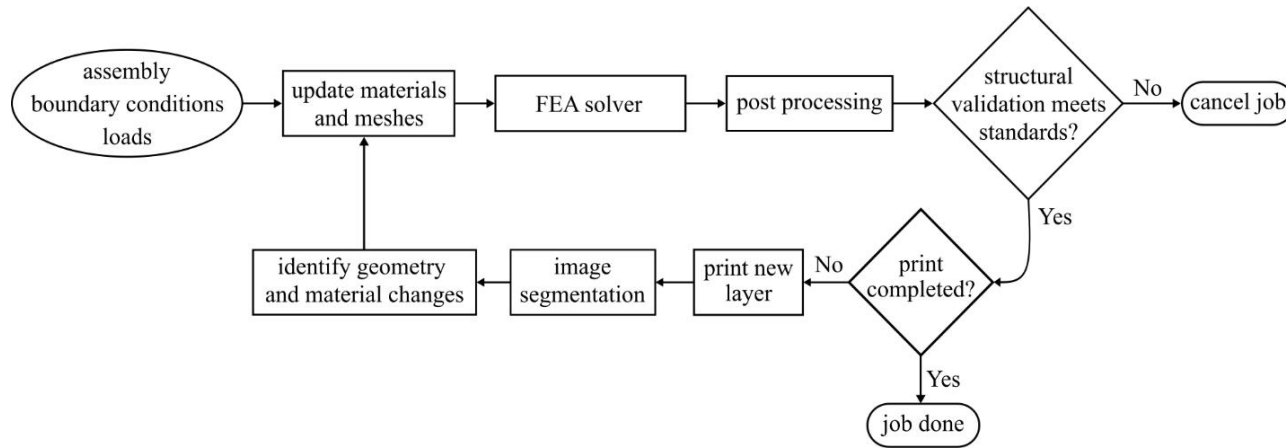
- **Decision Maker:** Uses DT insights to issue controls
 - Actions are directed towards the physical counterpart
 - Queries are directed towards the virtual representation
- **Query/Response Loop:** The virtual representation answers targeted questions and provides predictions for informed decisions



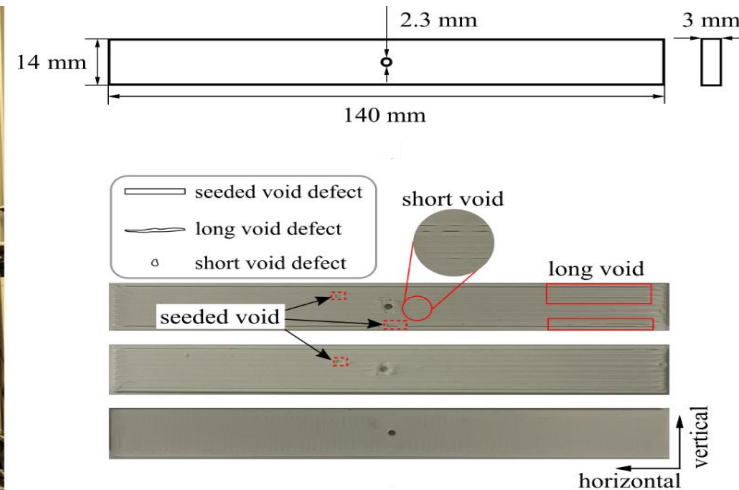
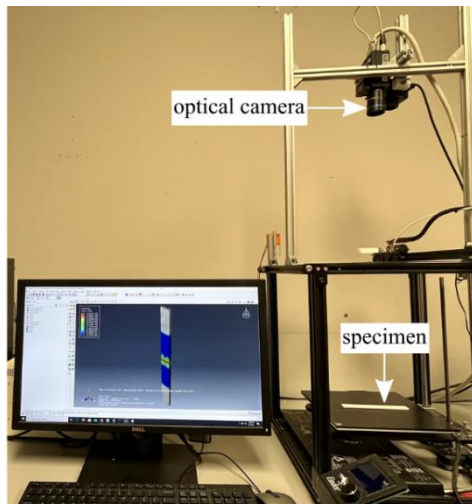
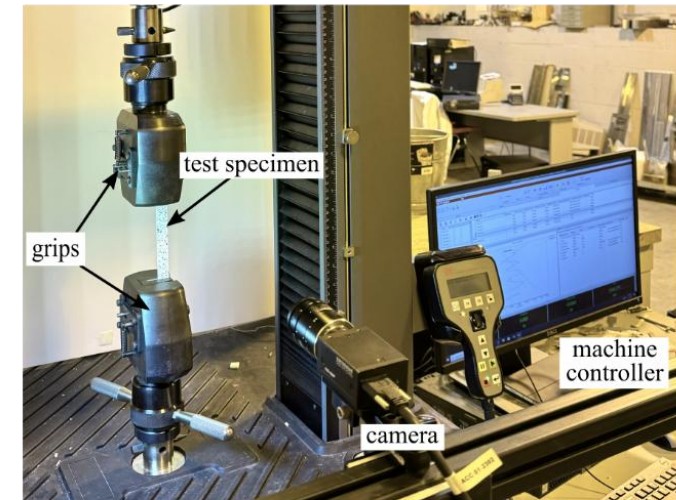
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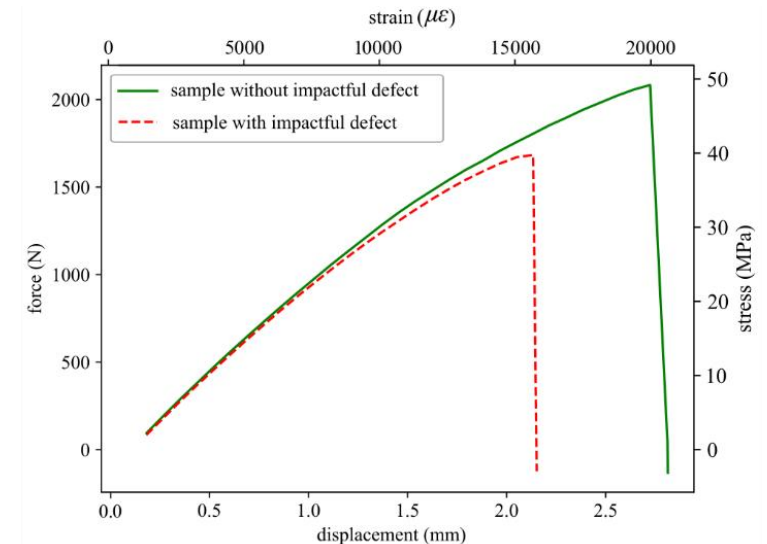
4. Model-based Simulation-in-the-loop Validation



flowchart of simulation-in-the-loop validation

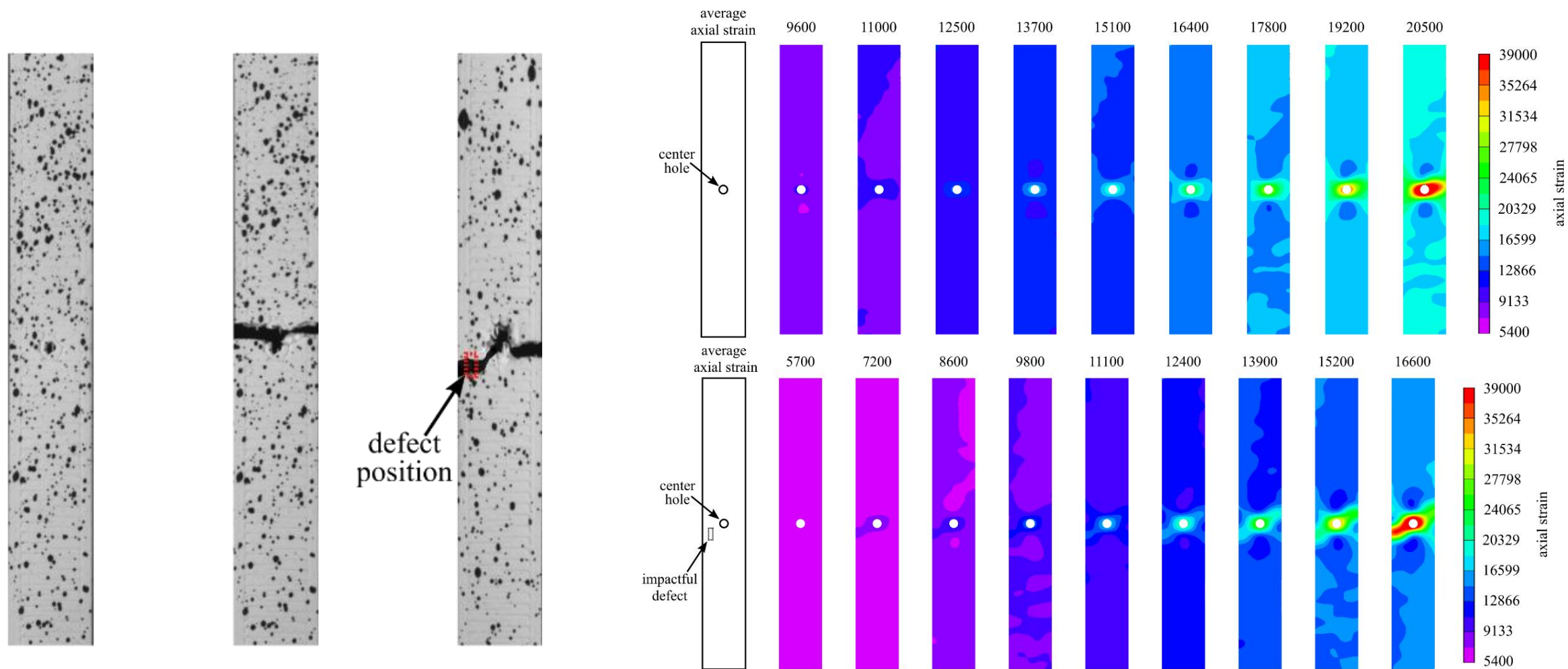


experiment setup and specimen with defect



experiment setup and specimen with defect

4. Model-based Simulation-in-the-loop Validation



speckled specimen

DIC results for specimen without and with defect

4. Model-based Simulation-in-the-loop Validation

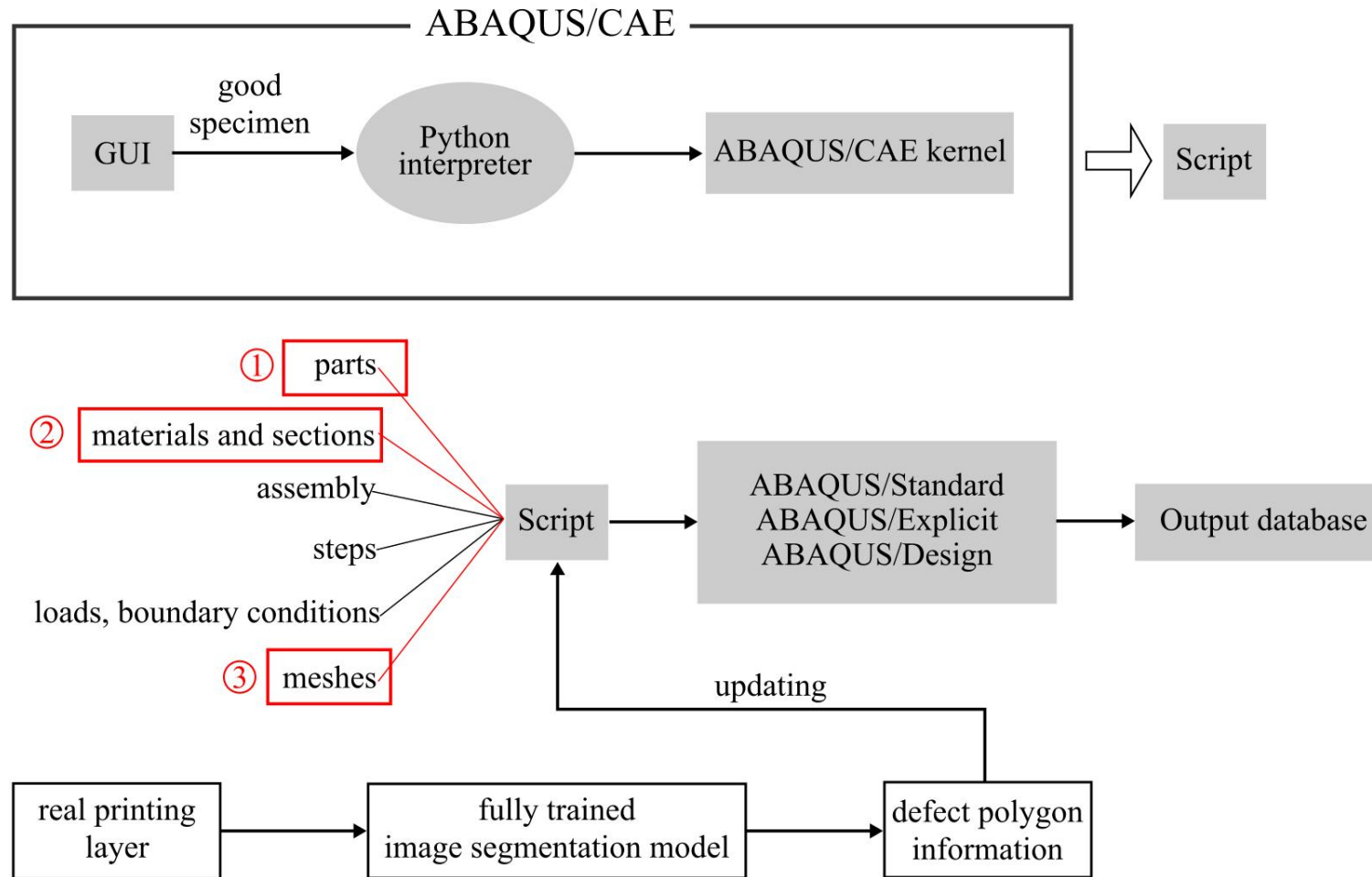
Table 1
Tensile and DIC test results with stress and strain difference.

Specimen	Test type	Number of samples	Averaged strain ($\mu\epsilon$)	Strain standard deviation	Strain difference ($\mu\epsilon$)	Averaged stress (MPa)	Stress standard deviation	Stress difference (MPa)
Without impactful defect	Tensile test	10	19 600	80.10	–	49.59	0.24	–
Without impactful defect	DIC test	10	20 500	86.32	900	50.61	0.28	1.02
With impactful defect	Tensile test	10	15 528	91.45	–	40.10	0.37	–
With impactful defect	DIC test	10	16 600	102.96	1072	41.43	0.43	1.33

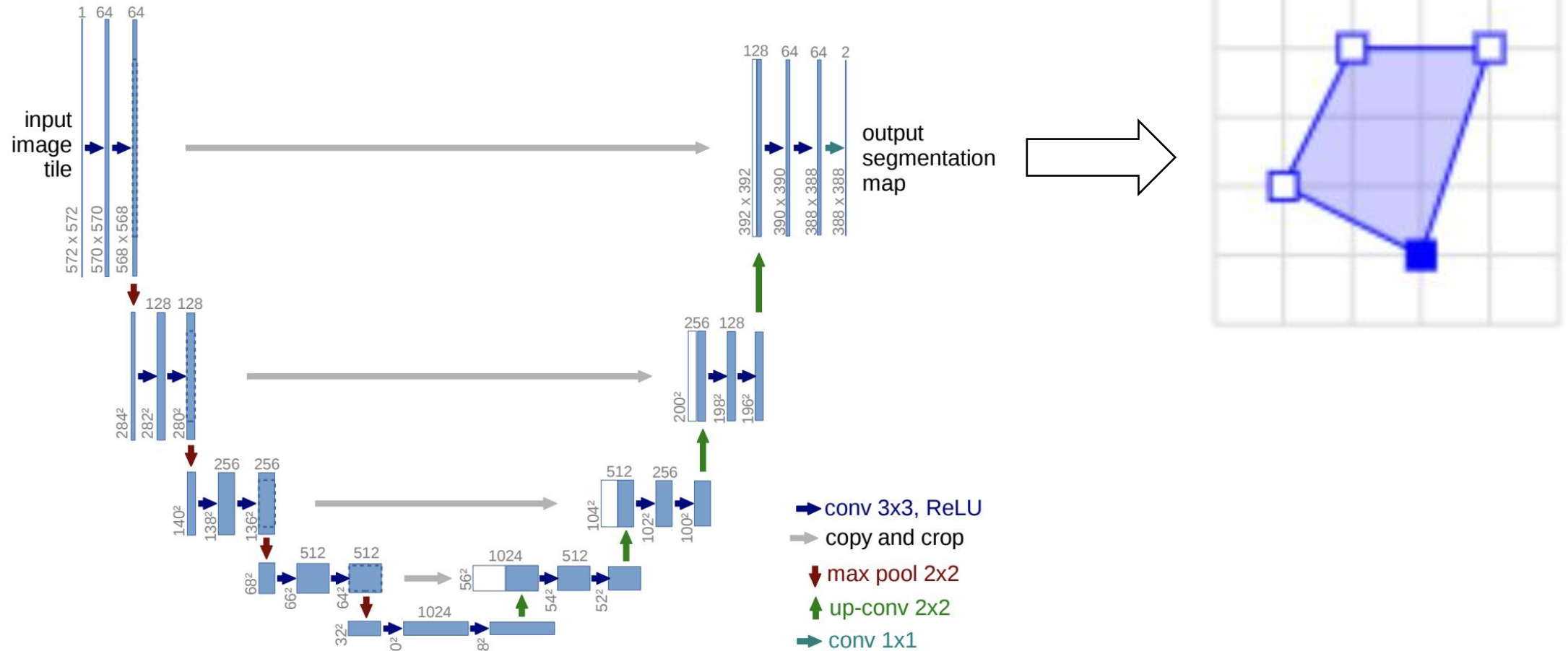
Table 2
Material property from tensile stress–strain for FEA.

Property	Density (g/mm^3)	Young’s modulus (MPa)	Poisson’s ratio	Strain at UTS ($\mu\epsilon$)	Ultimate Tensile Strength (UTS) (MPa)
Value	1.36	3000	0.36	19 600	49.59

4. Model-based Simulation-in-the-loop Validation



4. Model-based Simulation-in-the-loop Validation



4. Model-based Simulation-in-the-loop Validation

```
from part import *
from material import *
from assembly import *
from load import *
from mesh import *
from optimization import *
from job import *
from sketch import *
mdb.models['Model-1'].ConstrainedSketch(name='__profile__', sheetSize=200.0)
mdb.models['Model-1'].sketches['__profile__'].Line(point1=(0.0, 0.0), point2=(
0.0, 30.0))
mdb.models['Model-1'].sketches['__profile__'].Line(point1=(0.0, 30.0), point2=(
15.0, 30.0))
mdb.models['Model-1'].sketches['__profile__'].Line(point1=(15.0, 30.0), point2=
(15.0, -25.0))
mdb.models['Model-1'].Part(dimensionality=THREE_D, name='Dogboen_test', type=
DEFORMABLE_BODY)
mdb.models['Model-1'].parts['Dogboen_test'].BaseSolidExtrude(depth=5.0, sketch=
mdb.models['Model-1'].sketches['__profile__'])
mdb.models['Model-1'].Material(name='PLAMaterial')
mdb.models['Model-1'].materials['PLAMaterial'].Density(table=((7.8e-09, ), ))
mdb.models['Model-1'].materials['PLAMaterial'].Elastic(table=((210000.0, 0.3), ))
mdb.models['Model-1'].materials['PLAMaterial'].Plastic(table=((770.0, 0.0), (
800.0, 0.01), (830.0, 0.02), (850.0, 0.1), (900.0, 0.15), (930.0, 0.35), (
960.0, 1.0), (1050.0, 3.0)))
mdb.models['Model-1'].rootAssembly.Set(cells=
mdb.models['Model-1'].rootAssembly.instances['Dogboen_test-1'].cells.getSequenceFromMask(
(['#8 ]', ), ), name='Set-1')
mdb.models['Model-1'].EncastreBC(createStepName='Step-1', localCsys=None, name=
'BC-1', region=mdb.models['Model-1'].rootAssembly.sets['Set-1'])
mdb.models['Model-1'].parts['Dogboen_test'].generateMesh()
mdb.models['Model-1'].rootAssembly.regenerate()
mdb.Job(activateLoadBalancing=False, atTime=None, contactPrint=OFF)
mdb.jobs['TensileTest'].submit(consistencyChecking=OFF, datacheckJob=True)
```

Module import

Creat sketch object

Draw model sketch

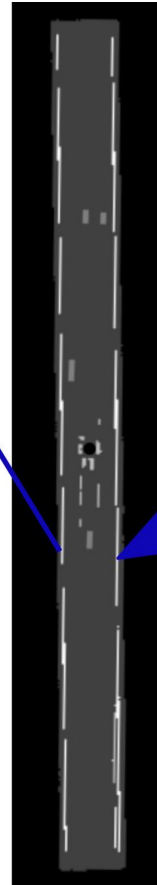
Creat model

Extrude sketch

Property setting

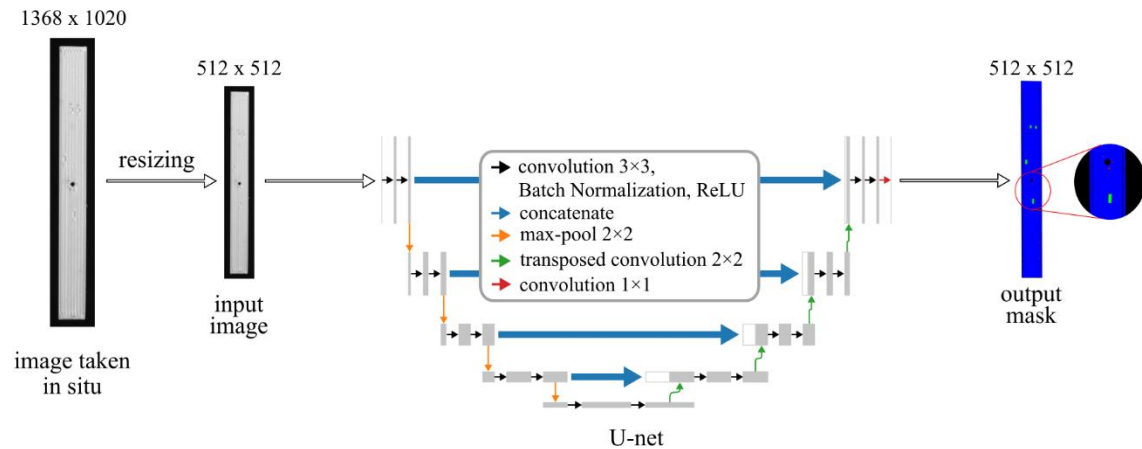
Boundary condition setting

Job submitting

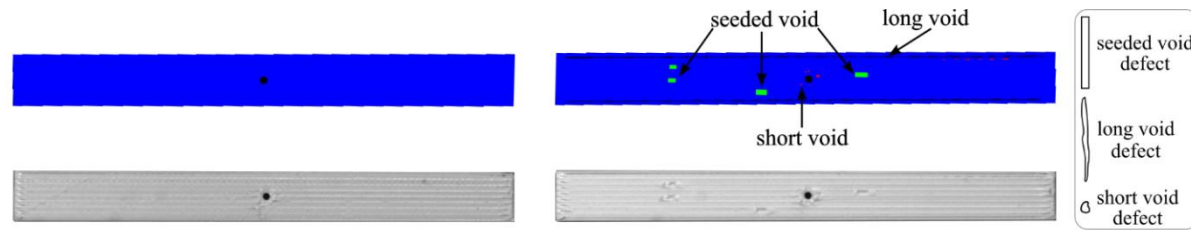


```
"annotations": [
{
  "id": 125686,
  "category_id": 2,
  "iscrowd": 0,
  "segmentation": [164.81, 417.51, 164.81, 444.46]
  "image_id": 242287,
  "area": 42061.80340000001,
  "bbox": [19.23, 383.18, 314.5, 244.46]
},
]
```

4. Model-based Simulation-in-the-loop Validation



U-net for defect segmentation



defect segmentation result

Manufacturing process

- layer-based

Image segmentation

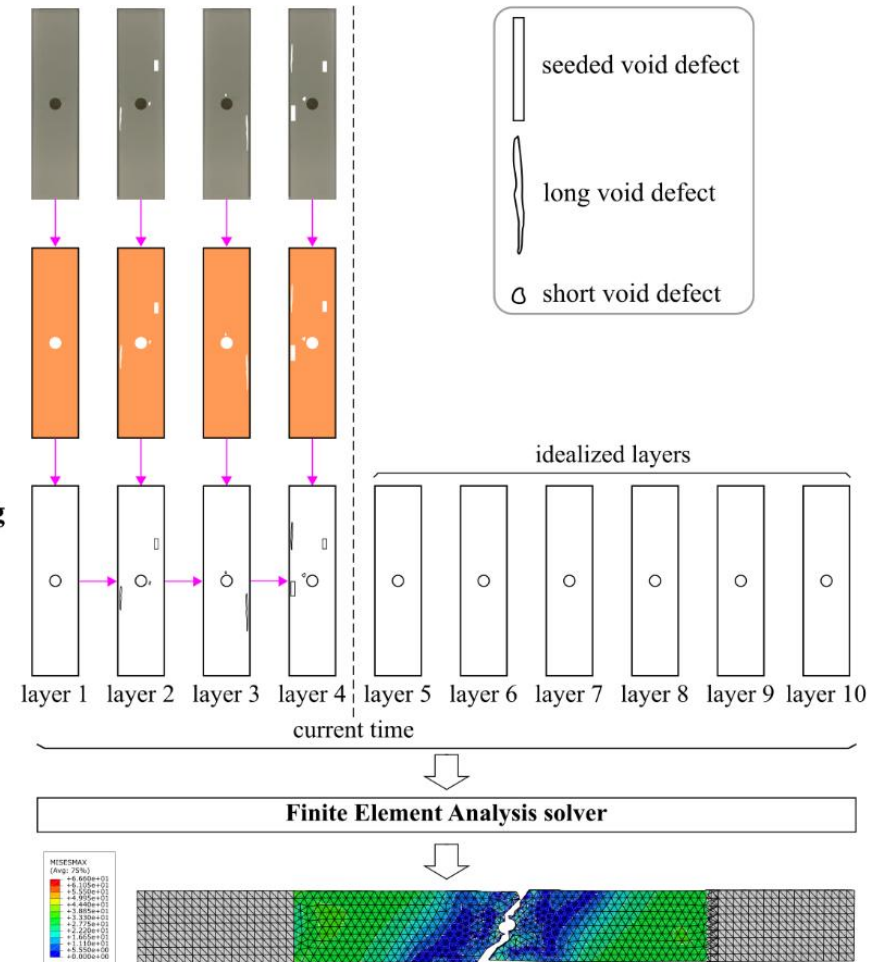
- U-net

Layer-based finite element model updating

- FEA script update
- void-related Young's modulus
- dynamic property for future layers

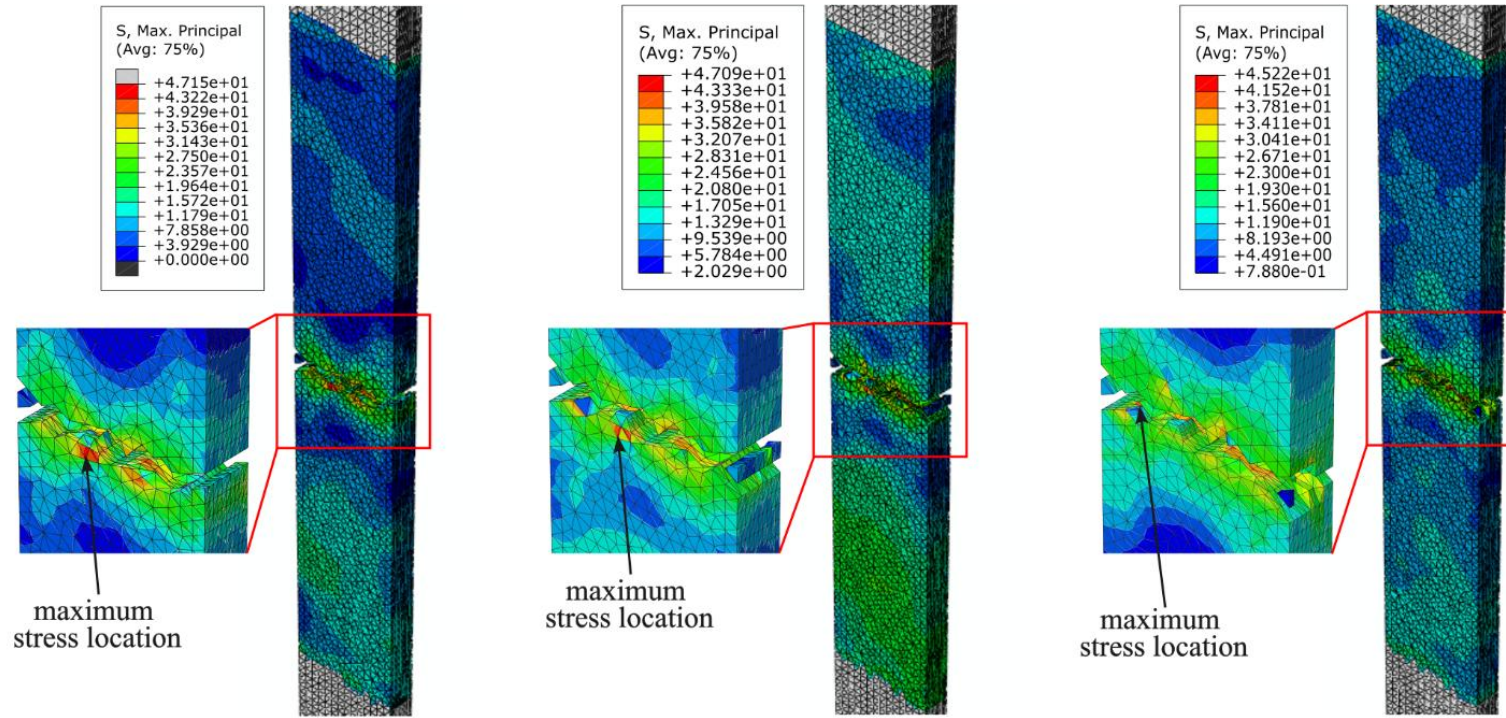
Online component structural validation

- structural validation
- decision making

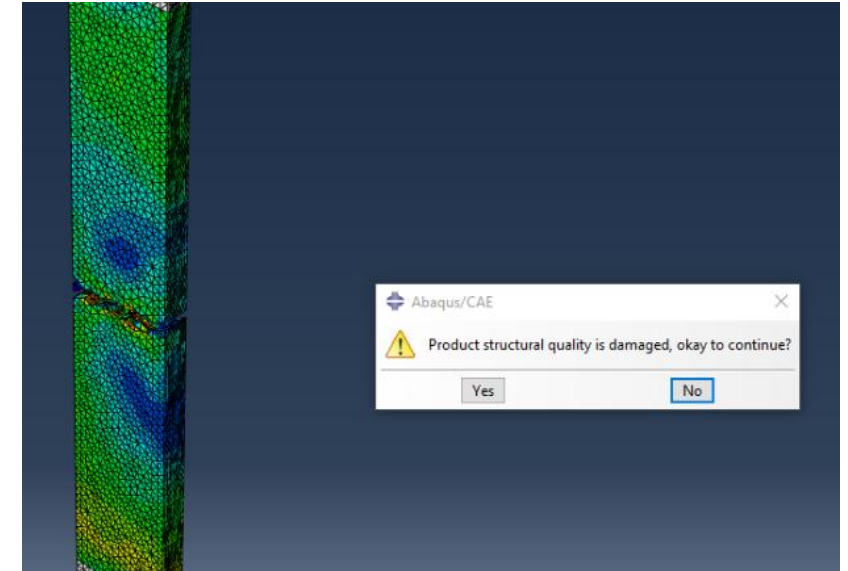


process of real-time simulation-in-the-loop validation

4. Model-based Simulation-in-the-loop Validation



FEA simulation results



cancelation alert

4. Model-based Simulation-in-the-loop Validation

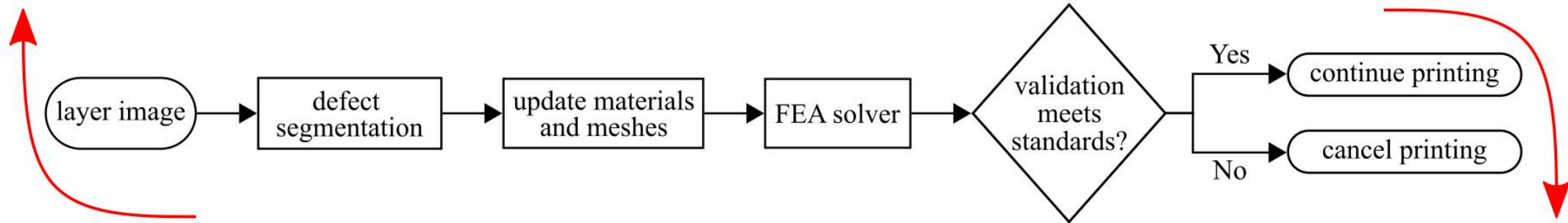
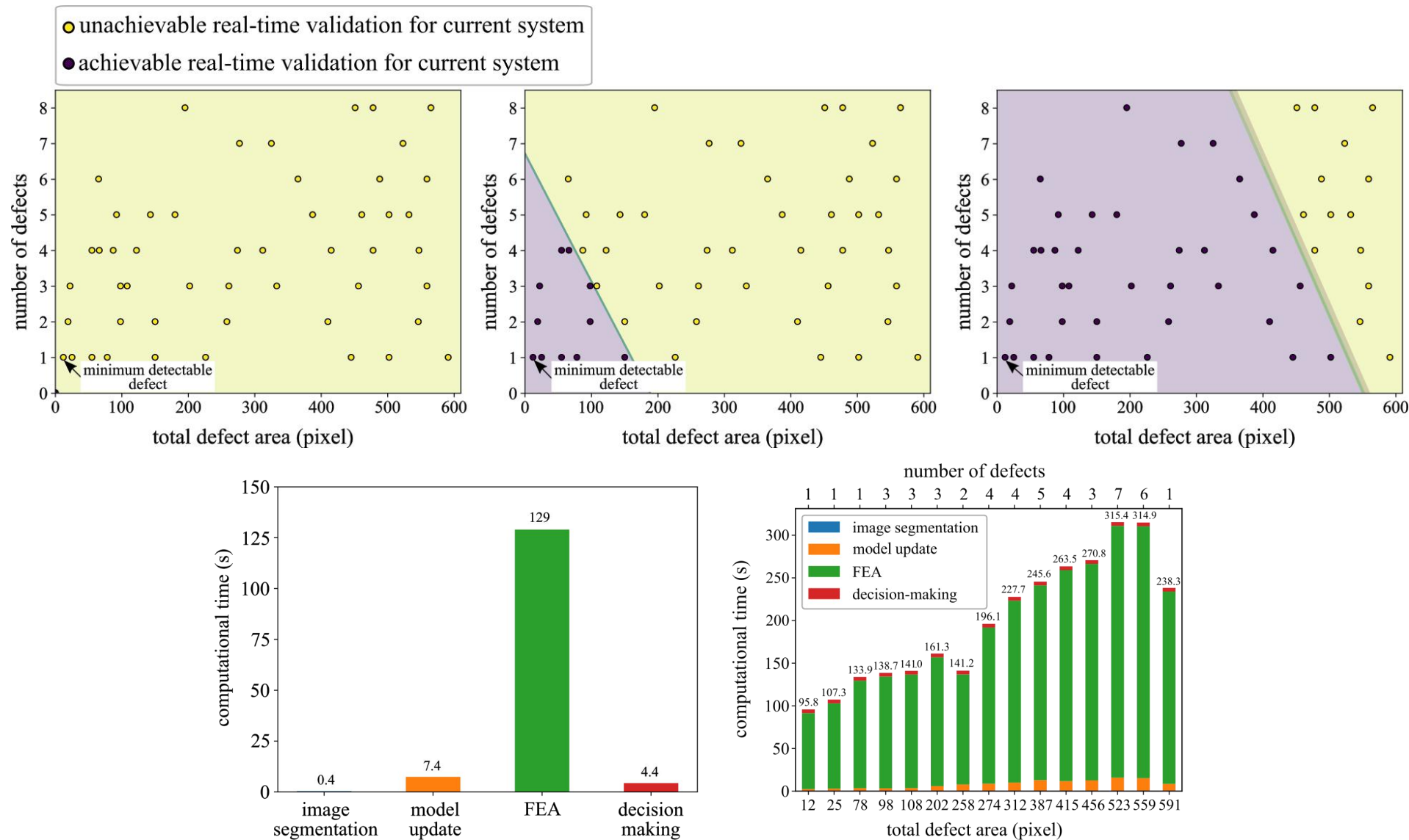


Table 3

System configurations for the real-time computational time calculation.

Hardware	System 1	System 2	System 3
Processor	Intel core i7-3770	AMD Ryzen Threadripper 3970X	Intel Xeon Gold 6250
Number of processor	1	1	2
Total core count	4 cores, 8 threads	32 cores, 64 threads	16 cores, 32 threads
Base clock speed	3.4 GHz	3.7 GHz	3.9 GHz
RAM	8 GB	128 GB	96 GB
Operating system	Windows 10, 64 Bit	Windows 10, 64 Bit	Windows 10, 64 Bit

4. Model-based Simulation-in-the-loop Validation



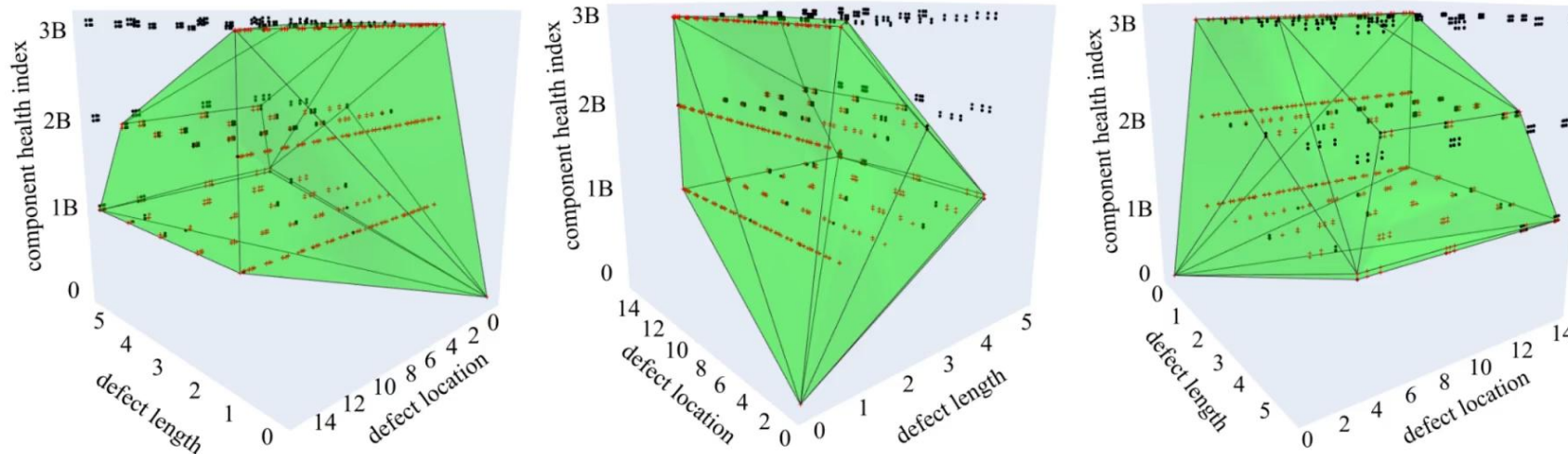
1. Fu, Y., Downey, A. R., Yuan, L., Huang, H. T., & Ogunniyi, E. A. (2025). *Additive Manufacturing*, 98, 104631.

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5. Data-driven Real-time Structural Validation

Real-time Product Structural Quality Validation for Material Extrusion Additive Manufacturing

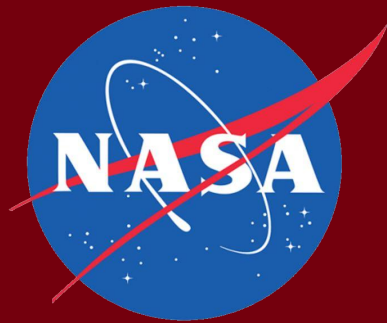


Discussion



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