

Who needs science to design materials?

Gareth Conduit

Theory of Condensed Matter Group, Department of Physics

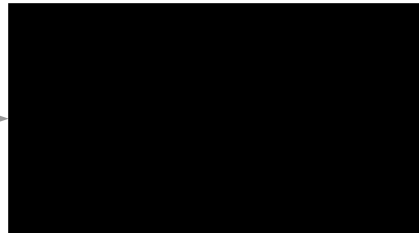
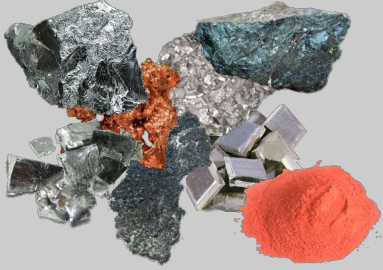
Neural network algorithm that can

Merge simulations, physical laws, and experimental data

Applications in materials discovery and drug design

Black box for materials design

Composition



Properties

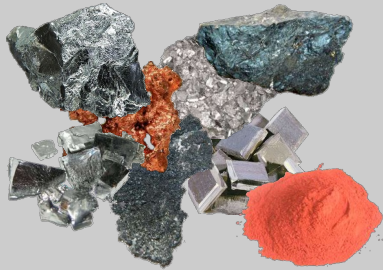
UTS

Hardness

Cost

Training the neural network

Composition

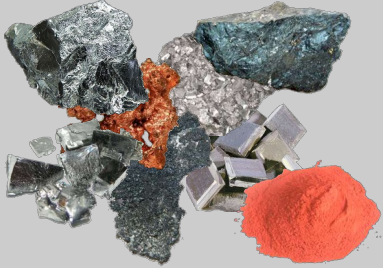


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597692868112392
376413439487341
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Properties
UTS
Hardness
Cost

Neural network for materials design

Composition



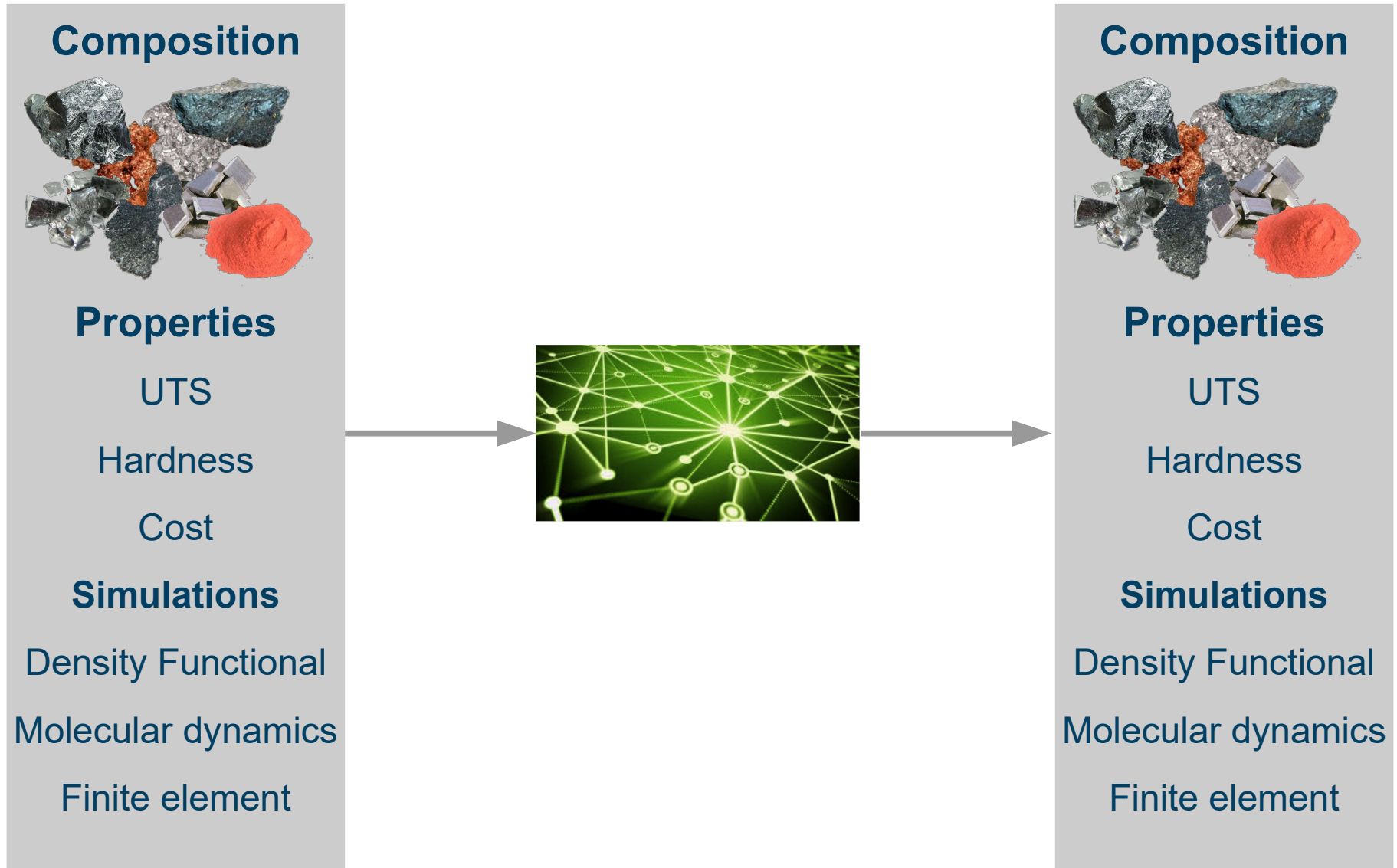
Properties

UTS

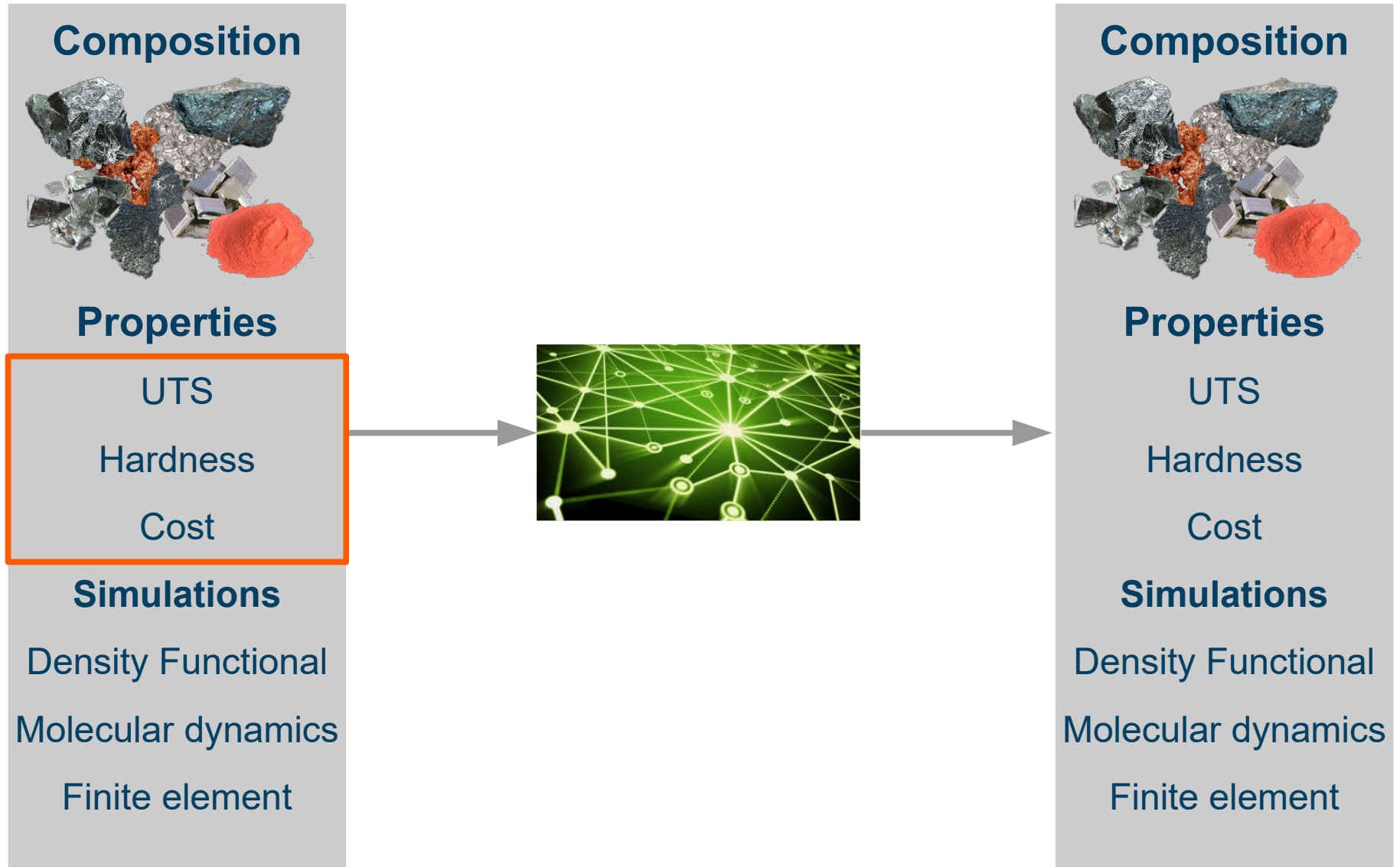
Hardness

Cost

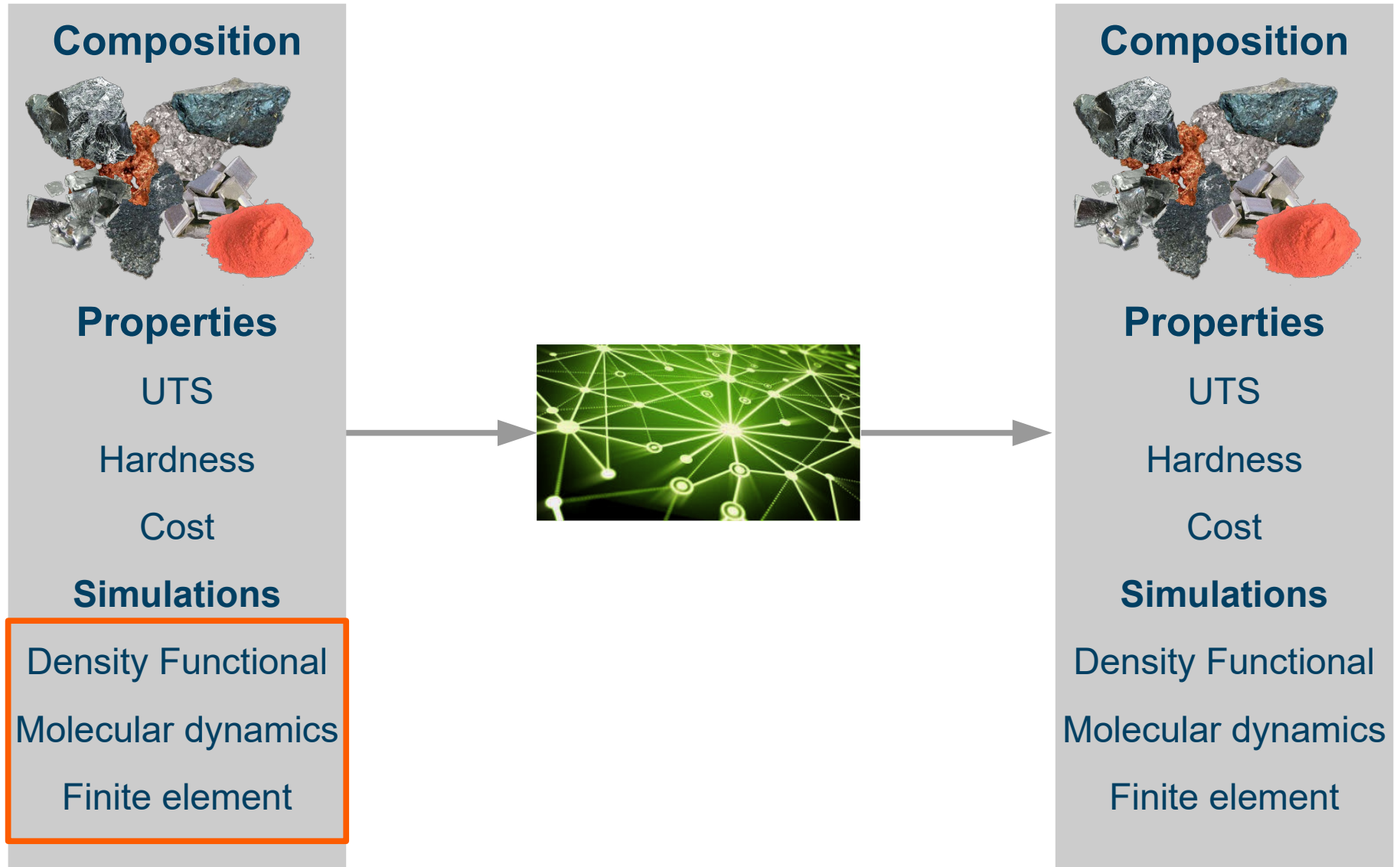
Neural network to exploit all available correlations



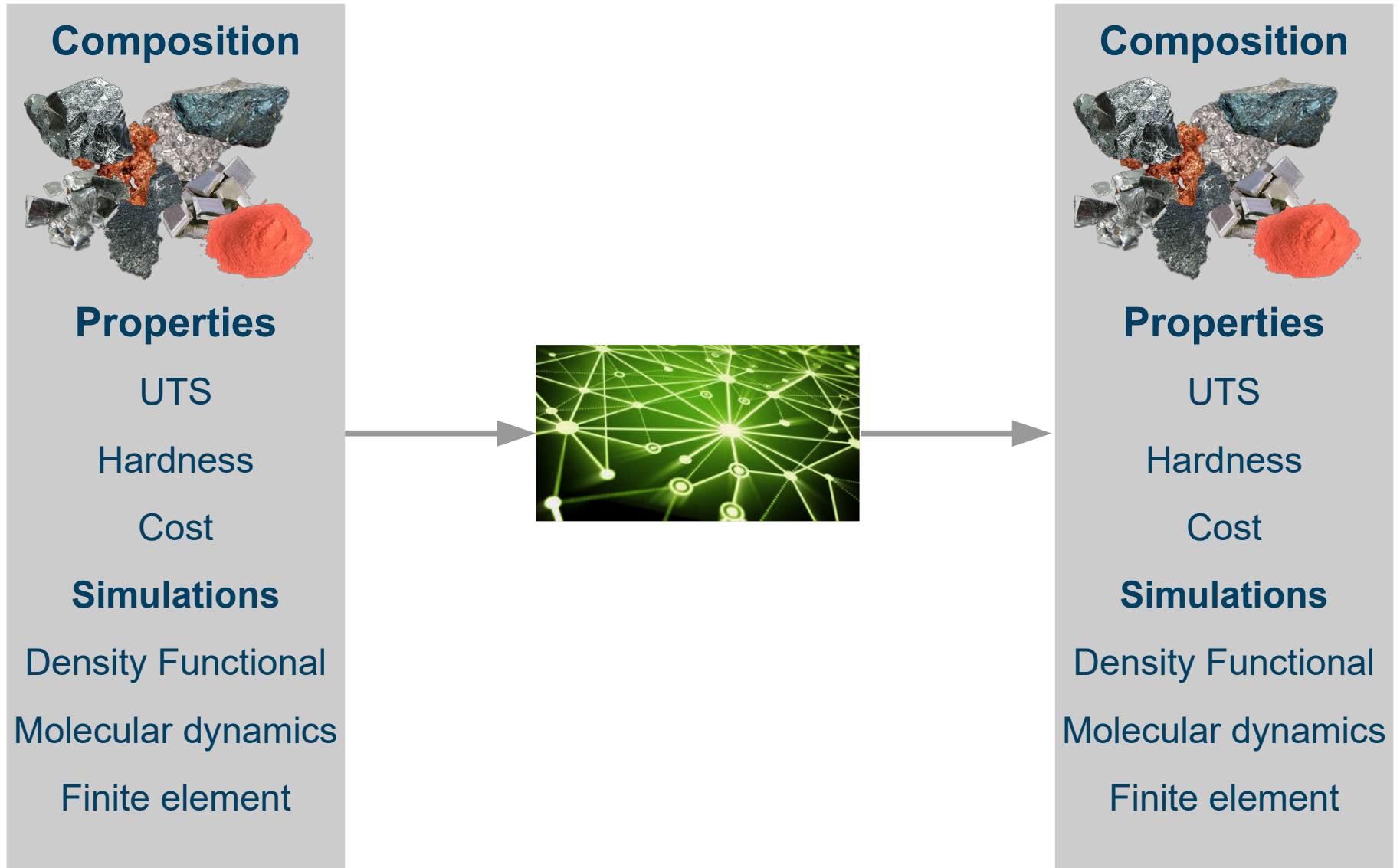
Neural network is top down



Neural network is top down and bottom up



Neural network must handle fragmented data



Fragmented training data set

Composition

UTS

Hardness



Neural network: train on complete data



Proposed neural network: train on fragmented data

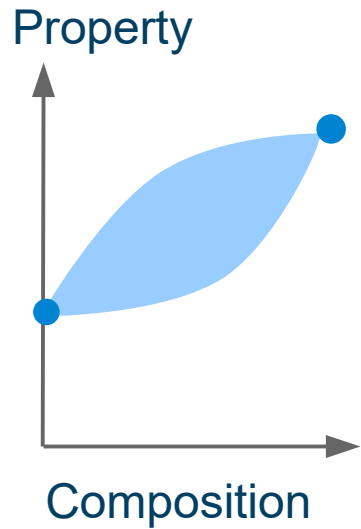


Proposed neural network: predict on fragmented data



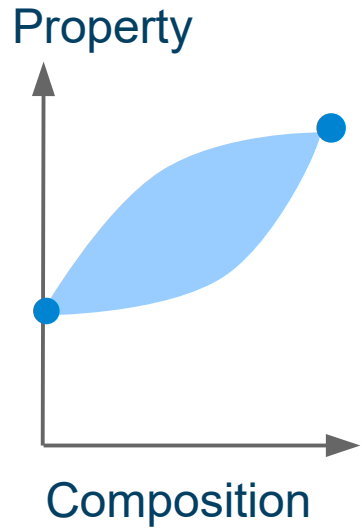
Neural network trained on experimental data

Experiment

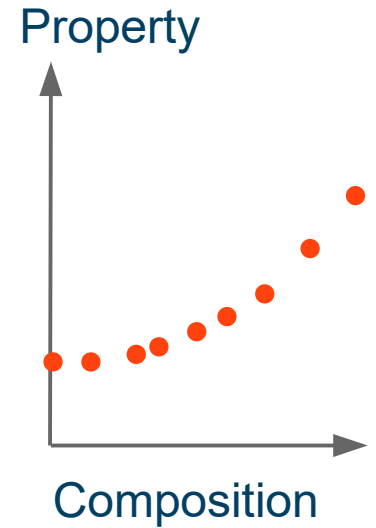


Further information is provided by a simulation

Experiment

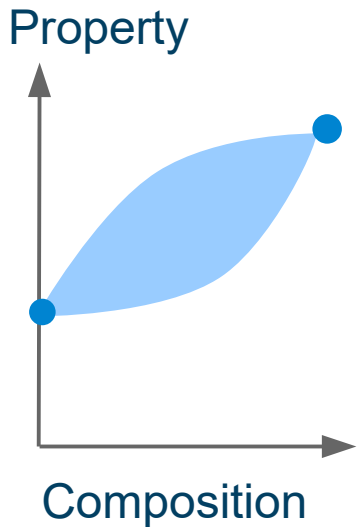


Simulation

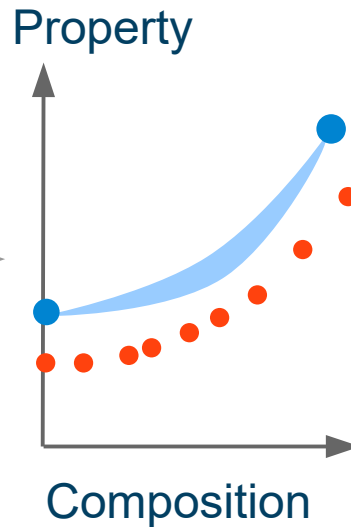


Neural network combines the two sources of data

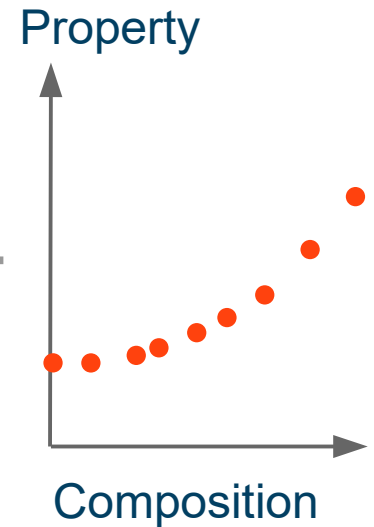
Experiment



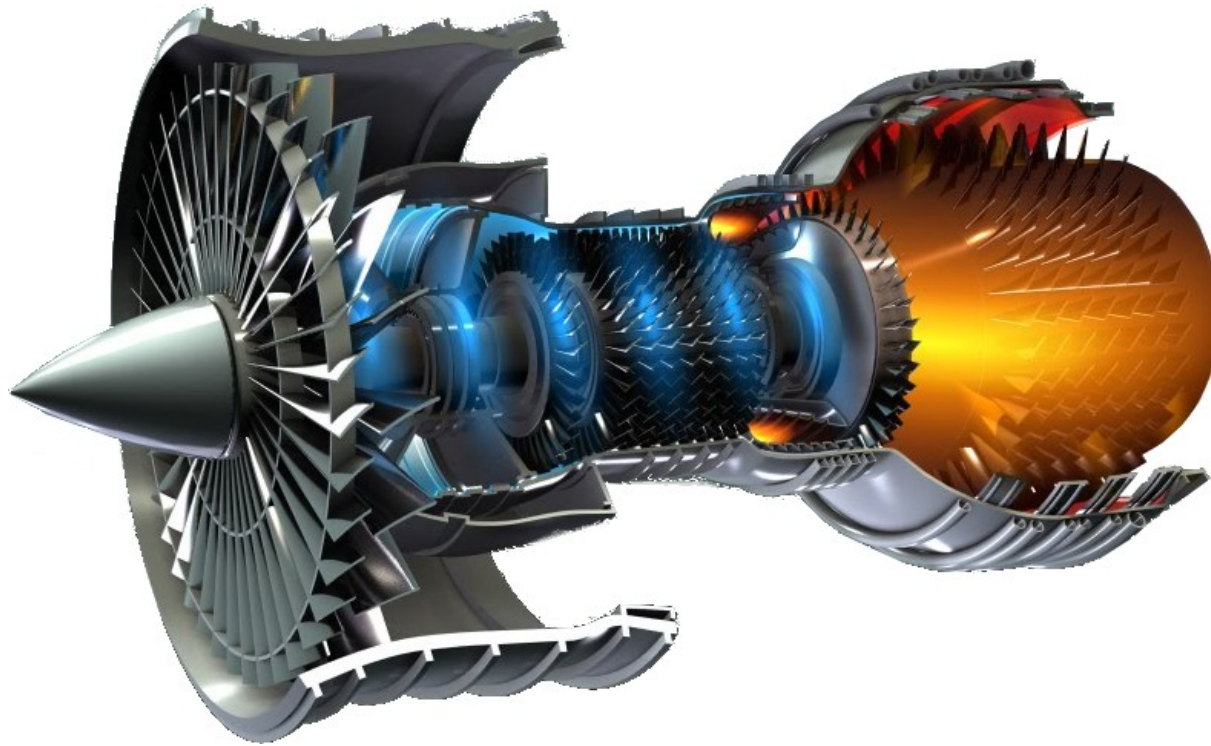
Combined



Simulation



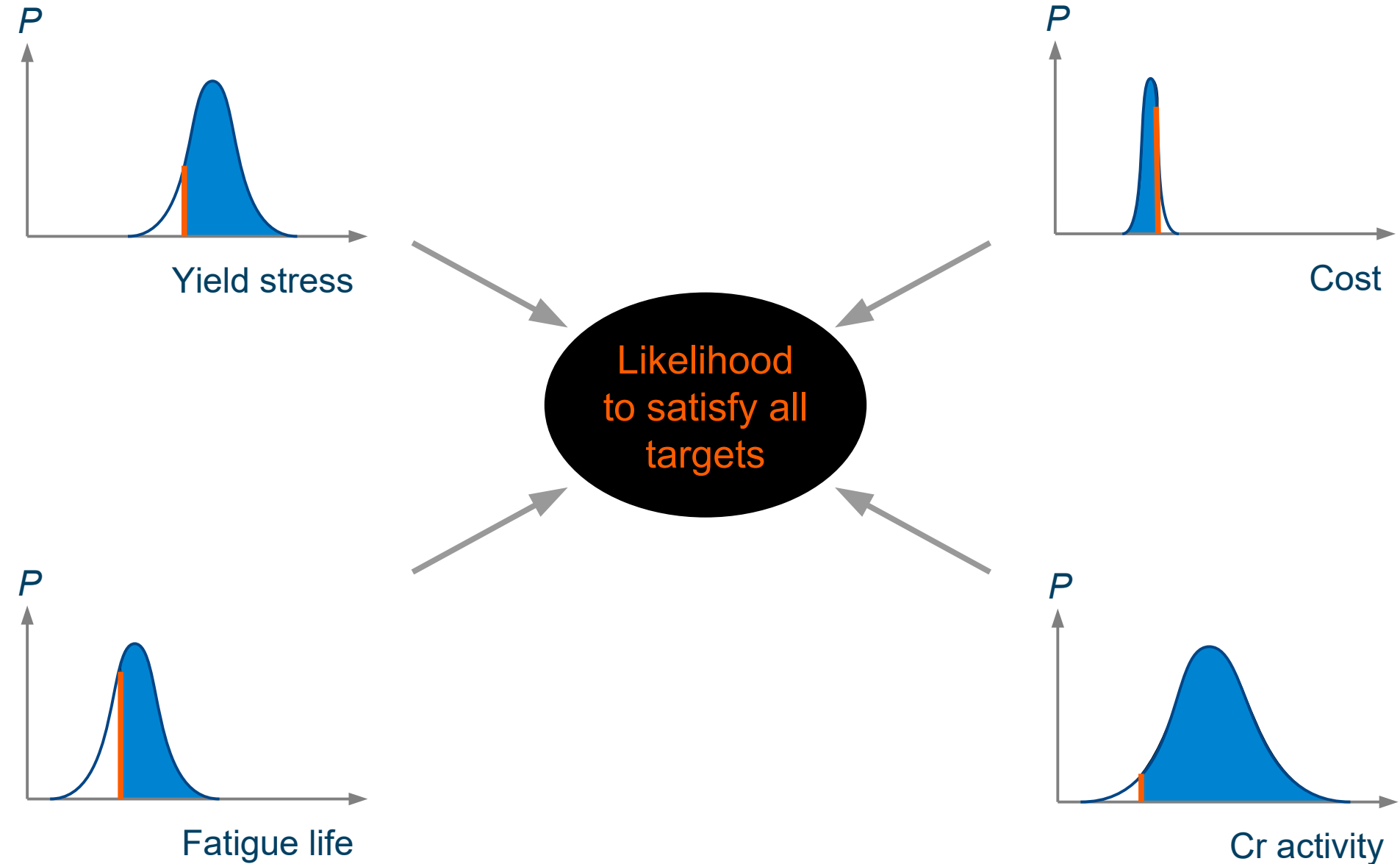
Schematic of an engine



Target properties

Cost	< 33.7 \$kg ⁻¹
Density	< 8281 kgm ⁻³
γ' content	< 50.4 vol%
Phase stability	> 99.0 vol%
Fatigue life	> 10 ^{3.9} cycles
Yield stress	> 752.2 MPa
Ultimate tensile strength	> 960.0 MPa
300hr stress rupture	> 674.5 MPa
Cr activity	> 0.14
γ' solvus	> 983°C
Tensile elongation	> 11.6%

Maximize the likelihood of success



Proposed alloy

Cr: 15.8



Co: 20.0



Mo: 0.5



W: 0.5



Ta: 4.9



Nb: 1.1



Al: 2.4



Ti: 3.0



Fe: 3.9



Mn: 0.2



Si: 0.2



C: 0.02



B: 0.06



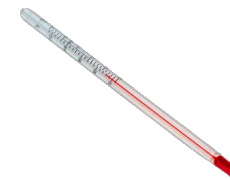
Zr: 0.18



Ni: 47.2



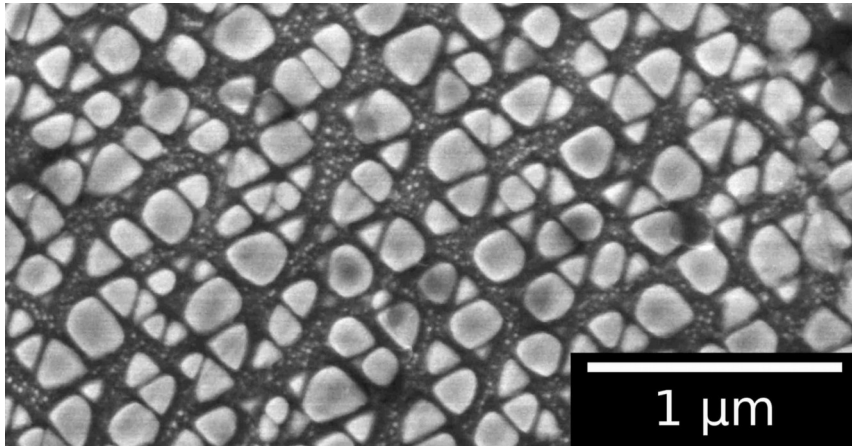
900°C



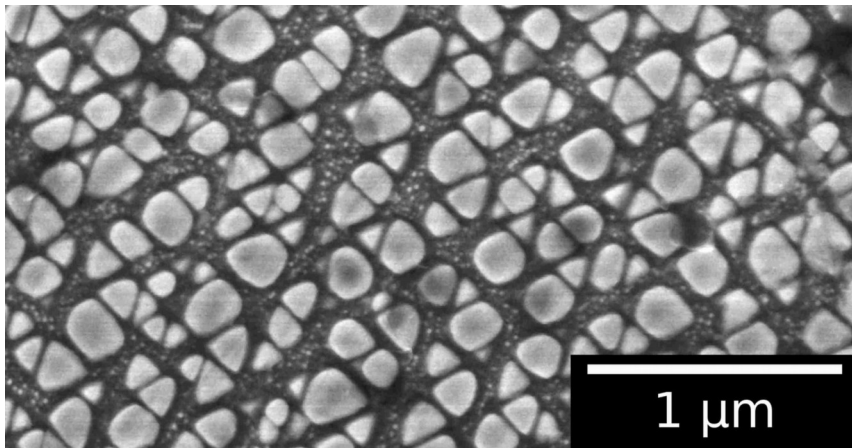
30 hours



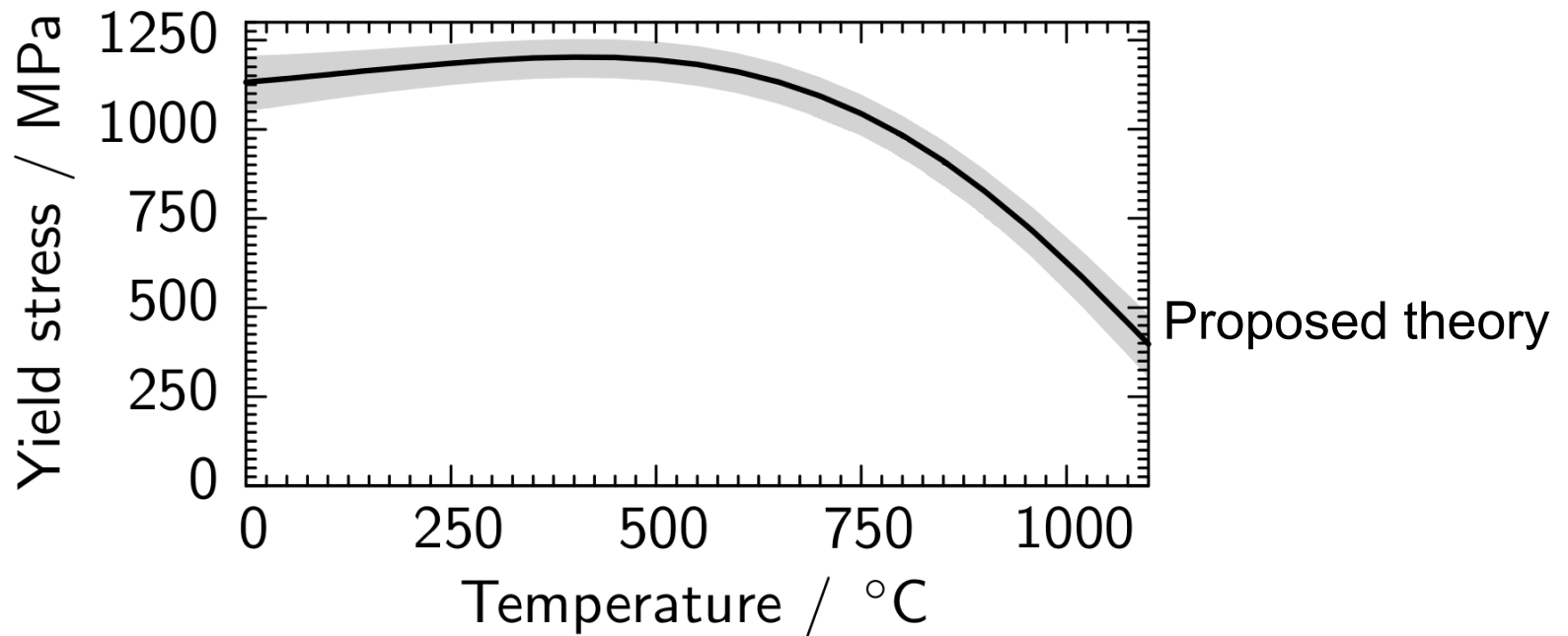
Microstructure



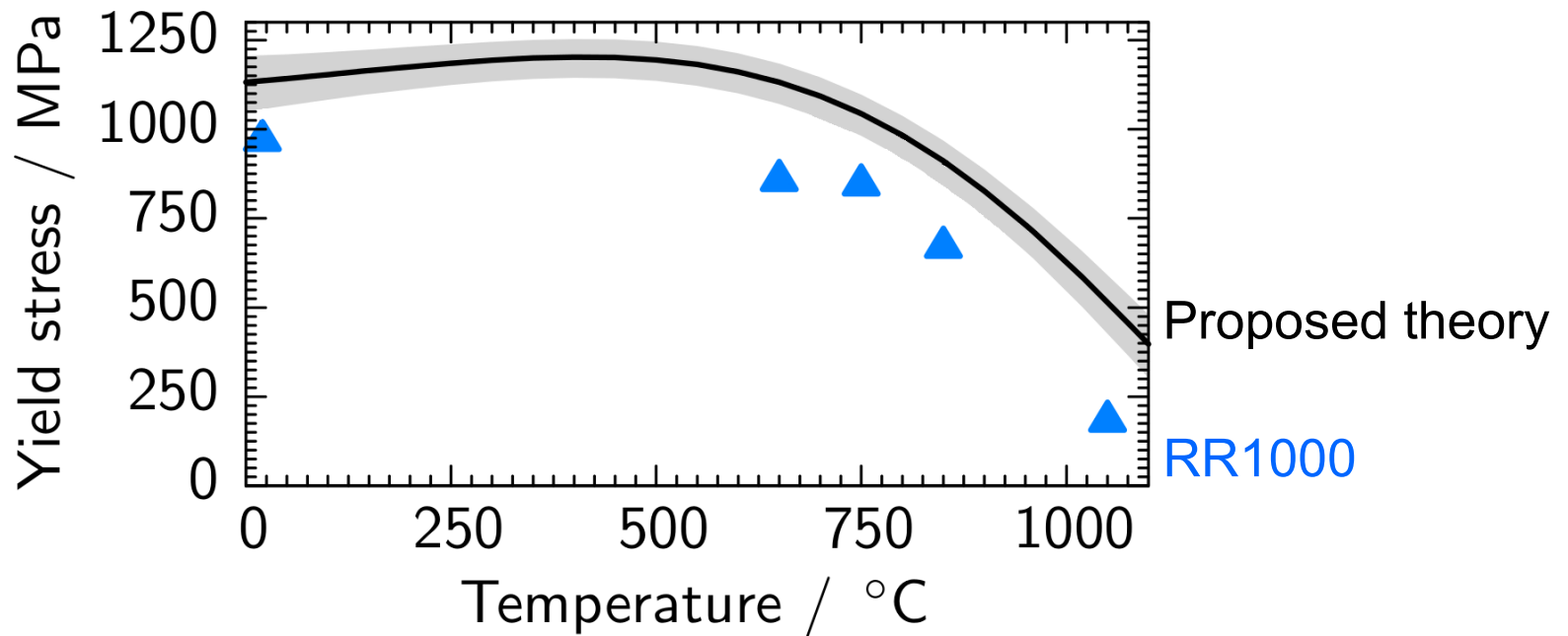
Precipitates strengthen the alloy



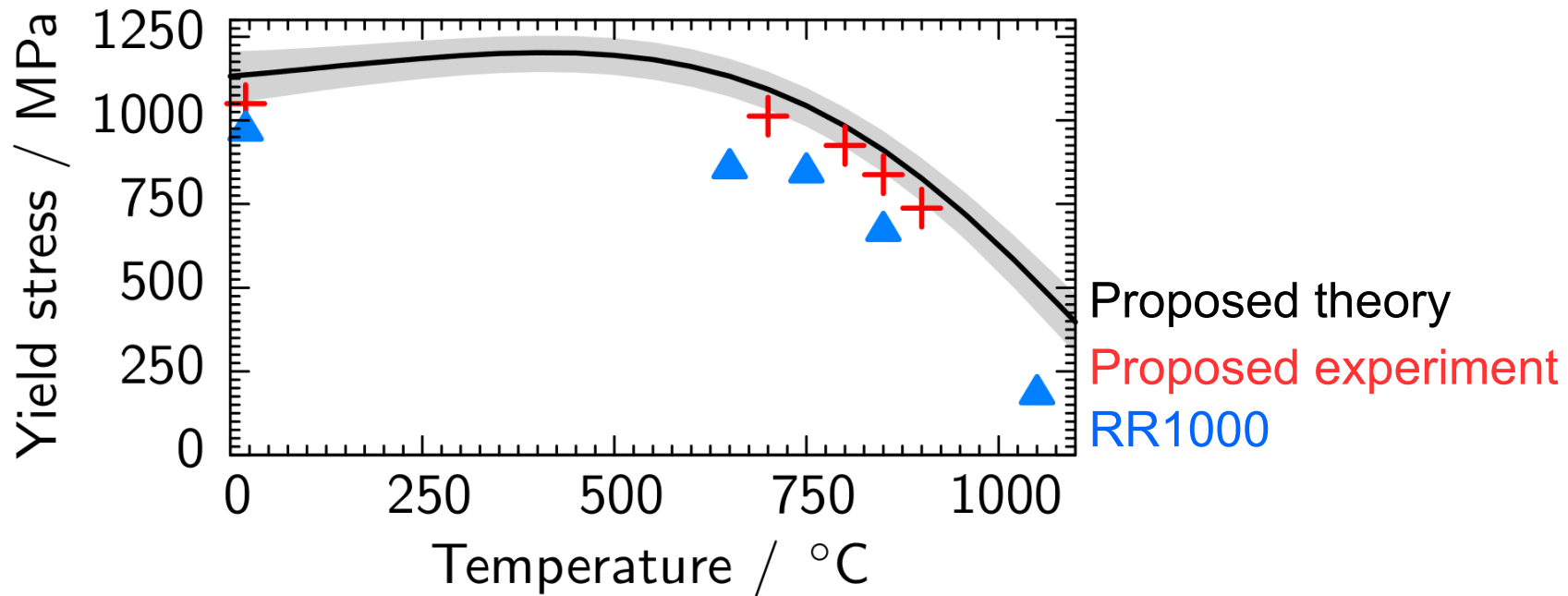
Predict the yield stress



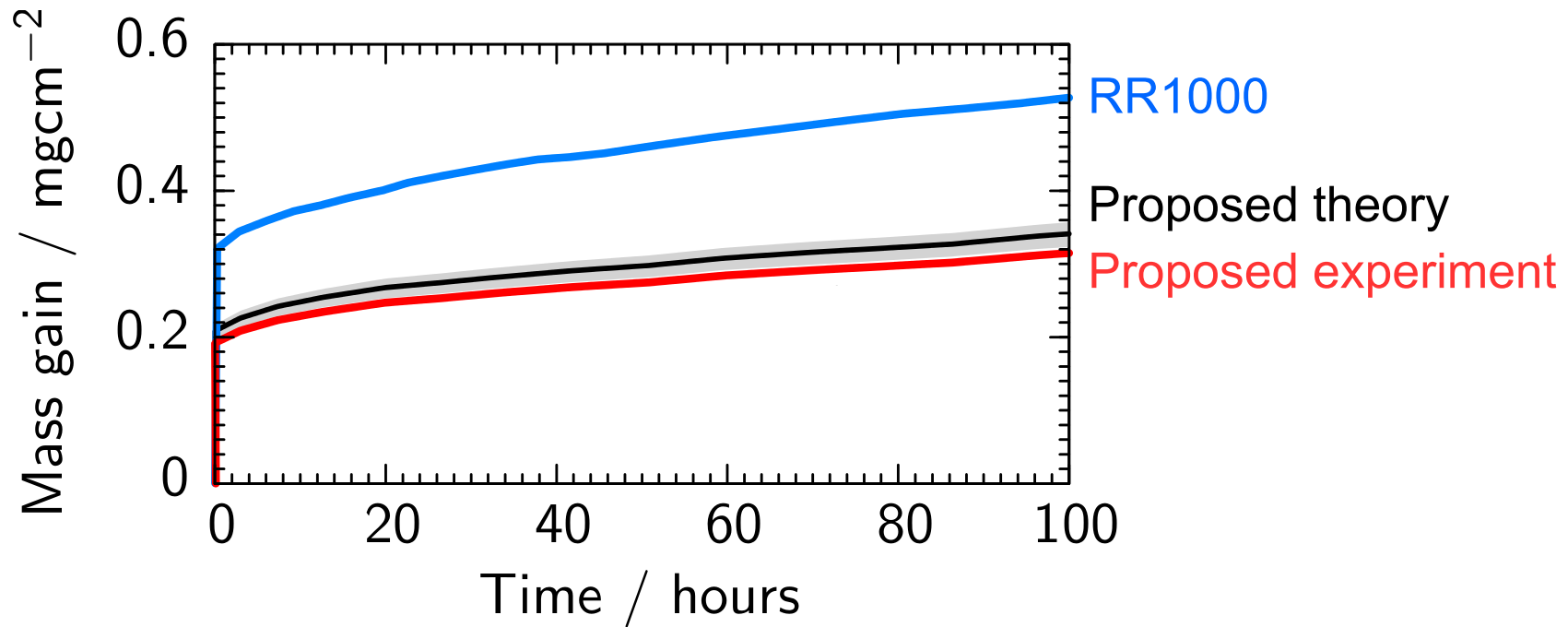
Test the yield stress



Test the yield stress

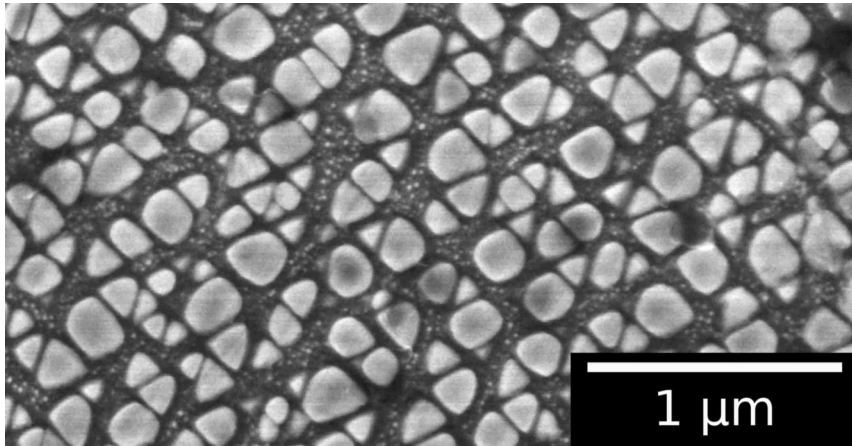


Test the oxidation resistance

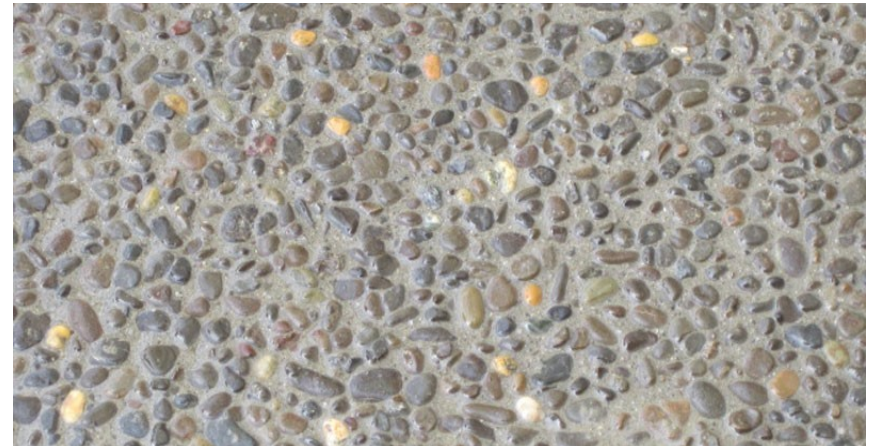


Microstructure strengthens the alloy

Precipitates

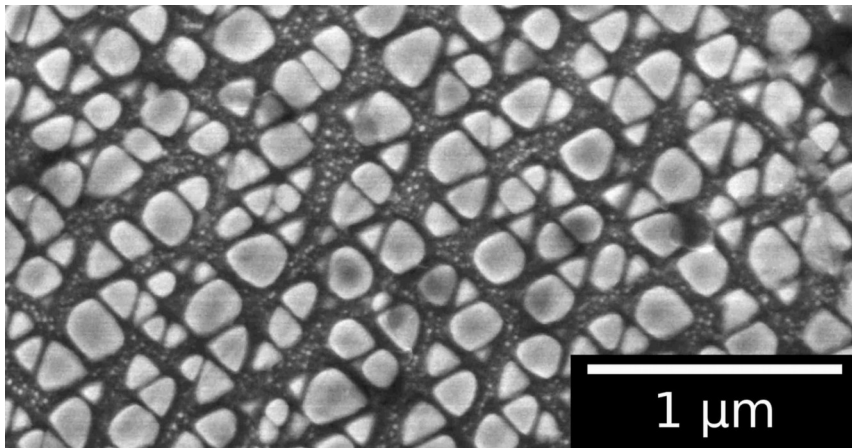


Aggregate

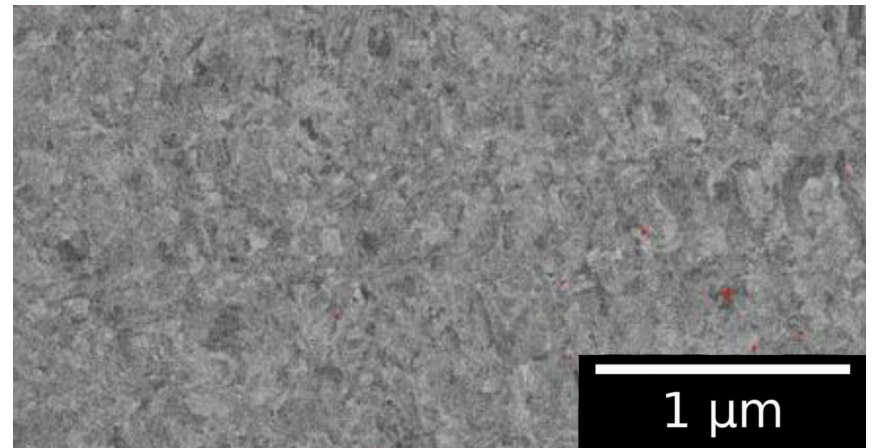


Microstructure defined by the heat treatment

Correct heat treatment

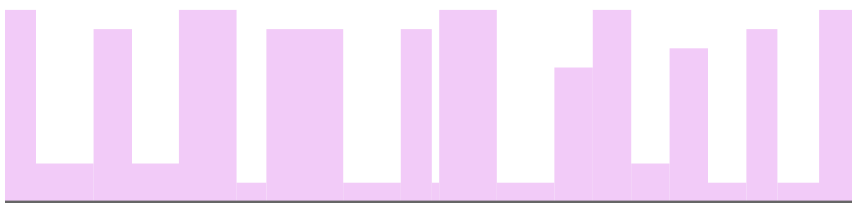
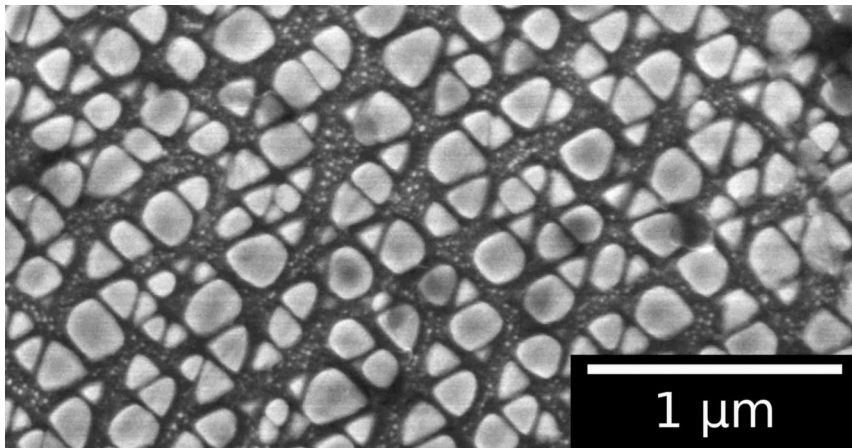


Incorrect heat treatment

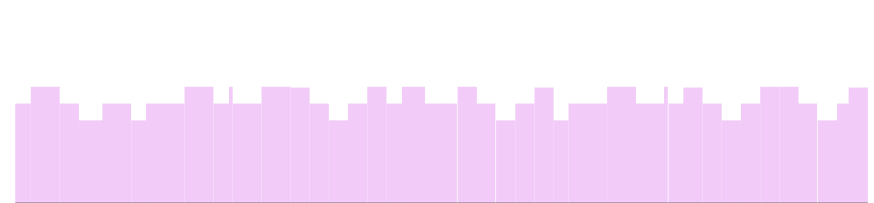
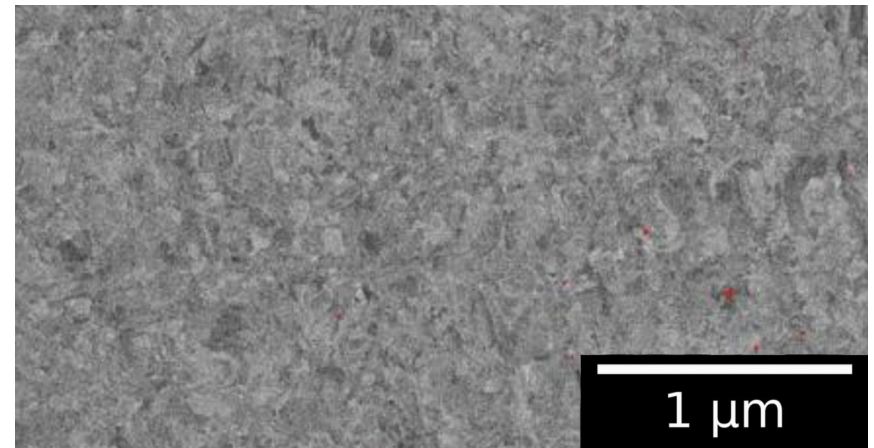


Microstructure noise contains information about the heat treatment

Correct heat treatment

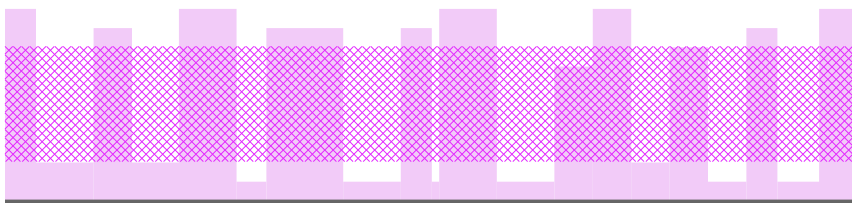
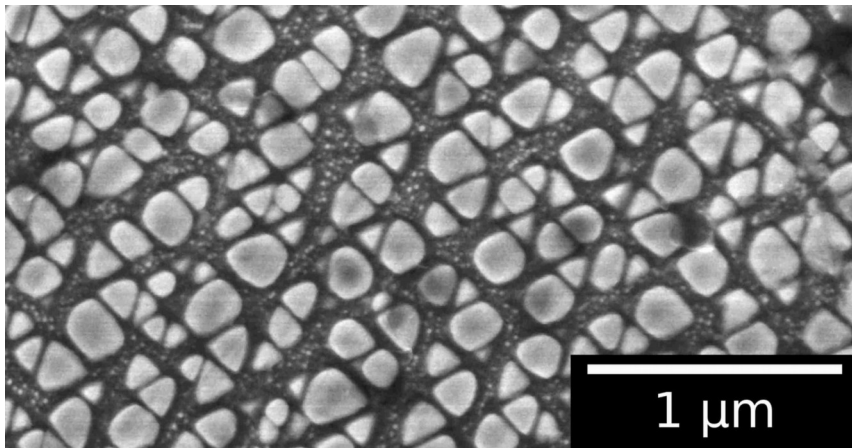


Incorrect heat treatment

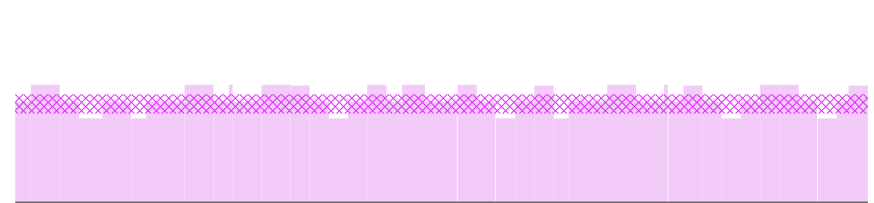
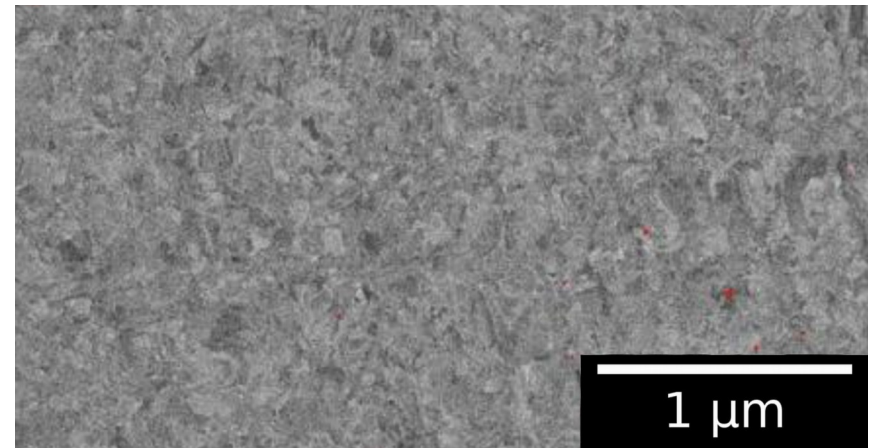


Microstructure noise contains information about the heat treatment

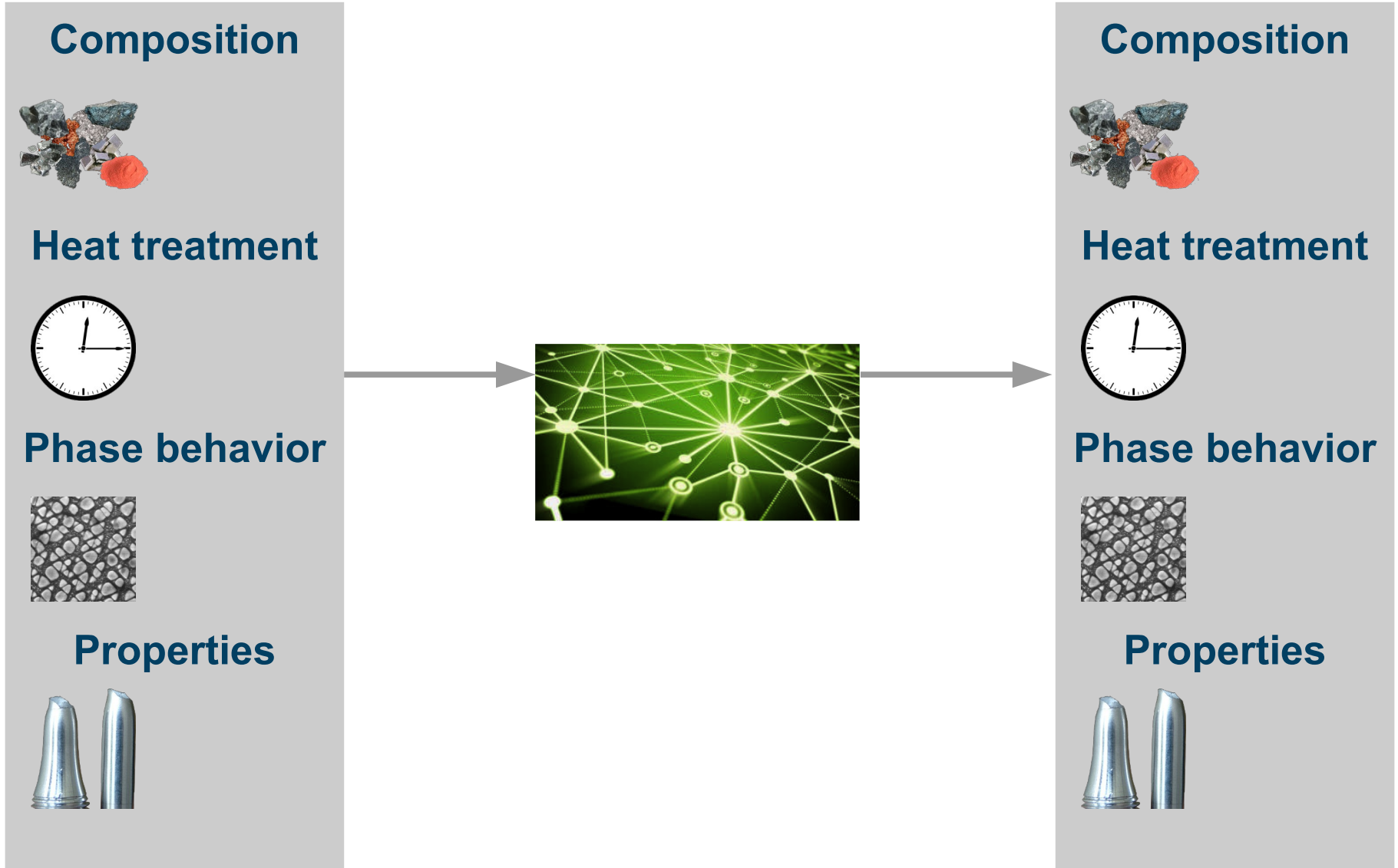
Correct heat treatment



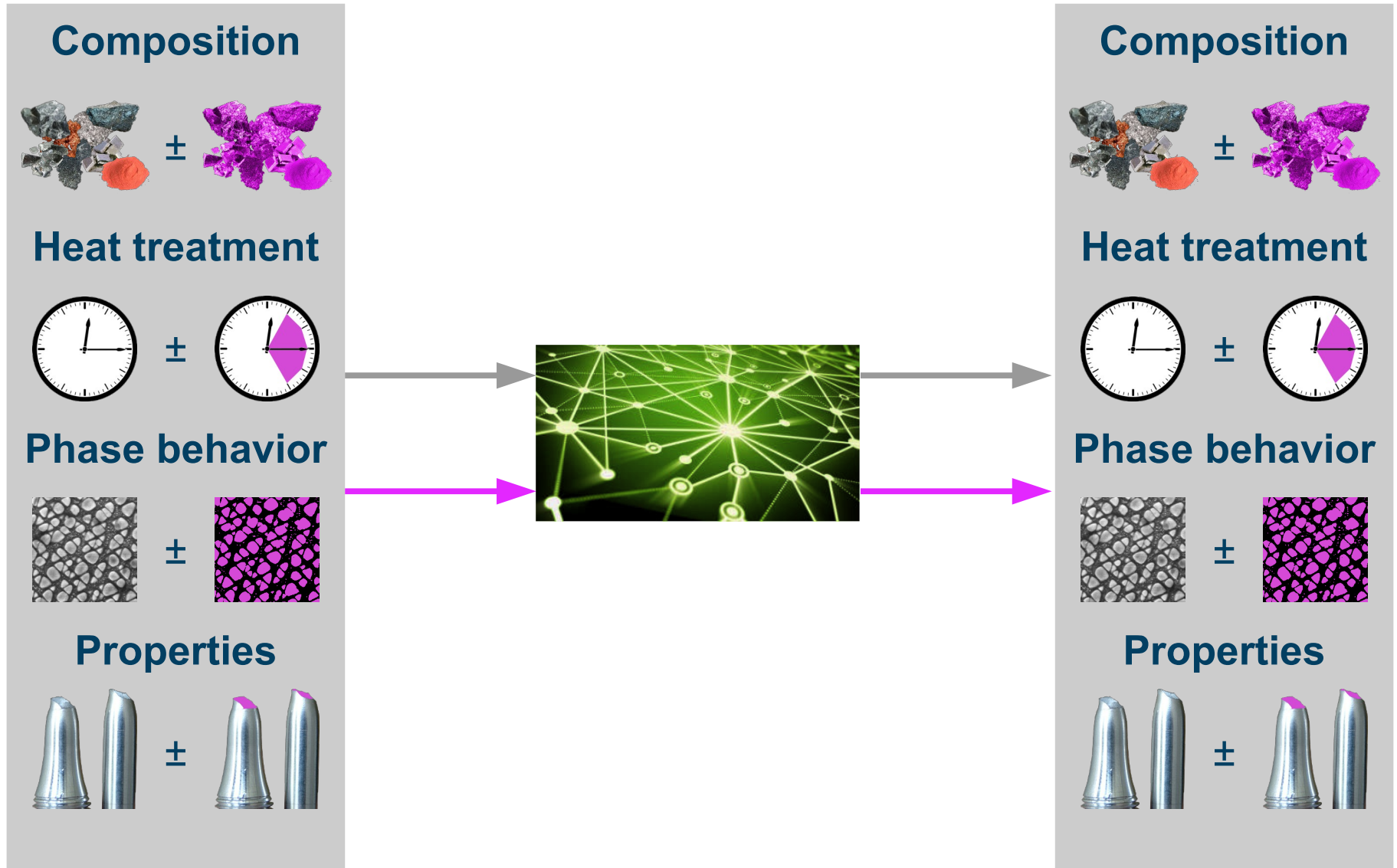
Incorrect heat treatment



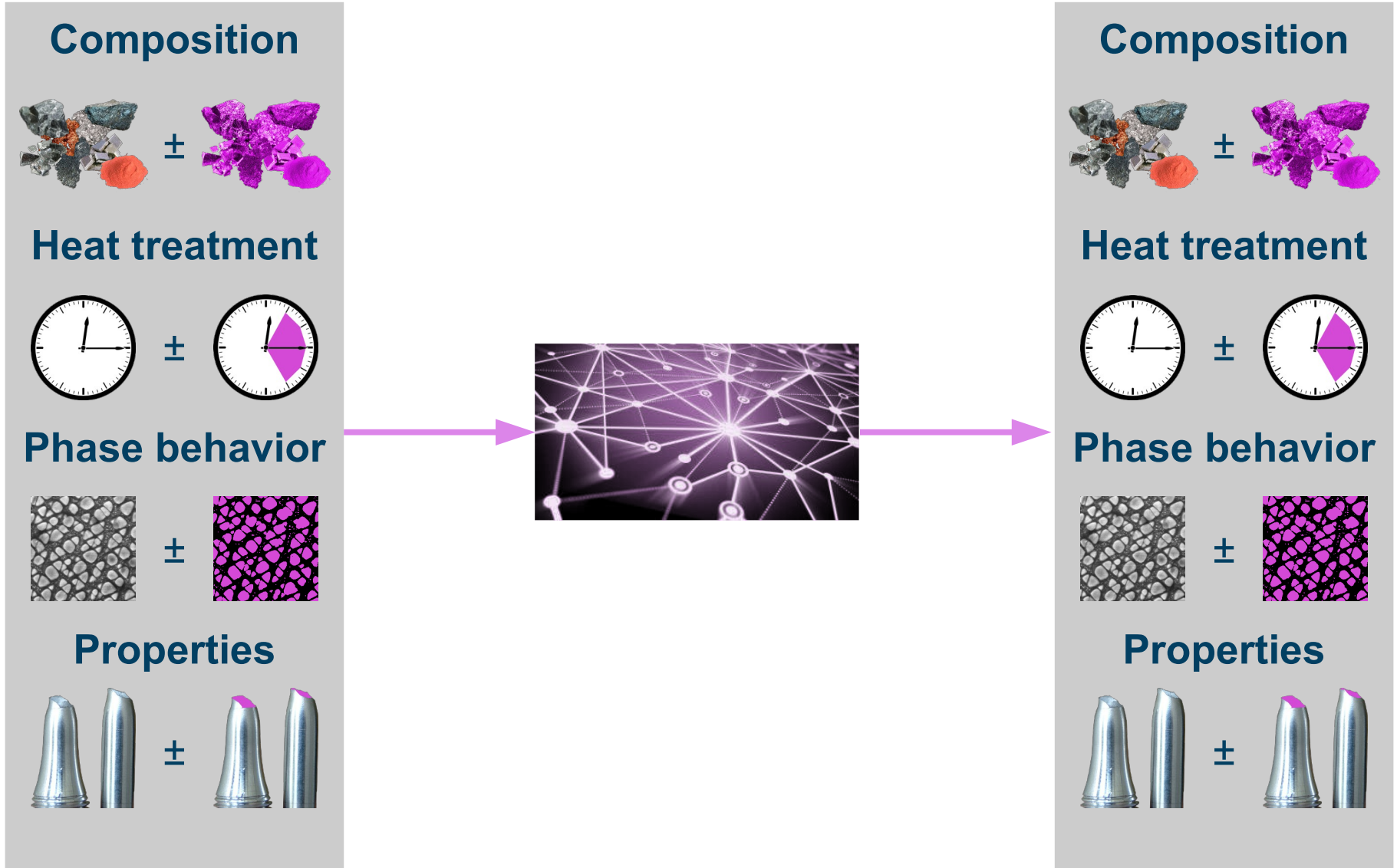
Standard neural network



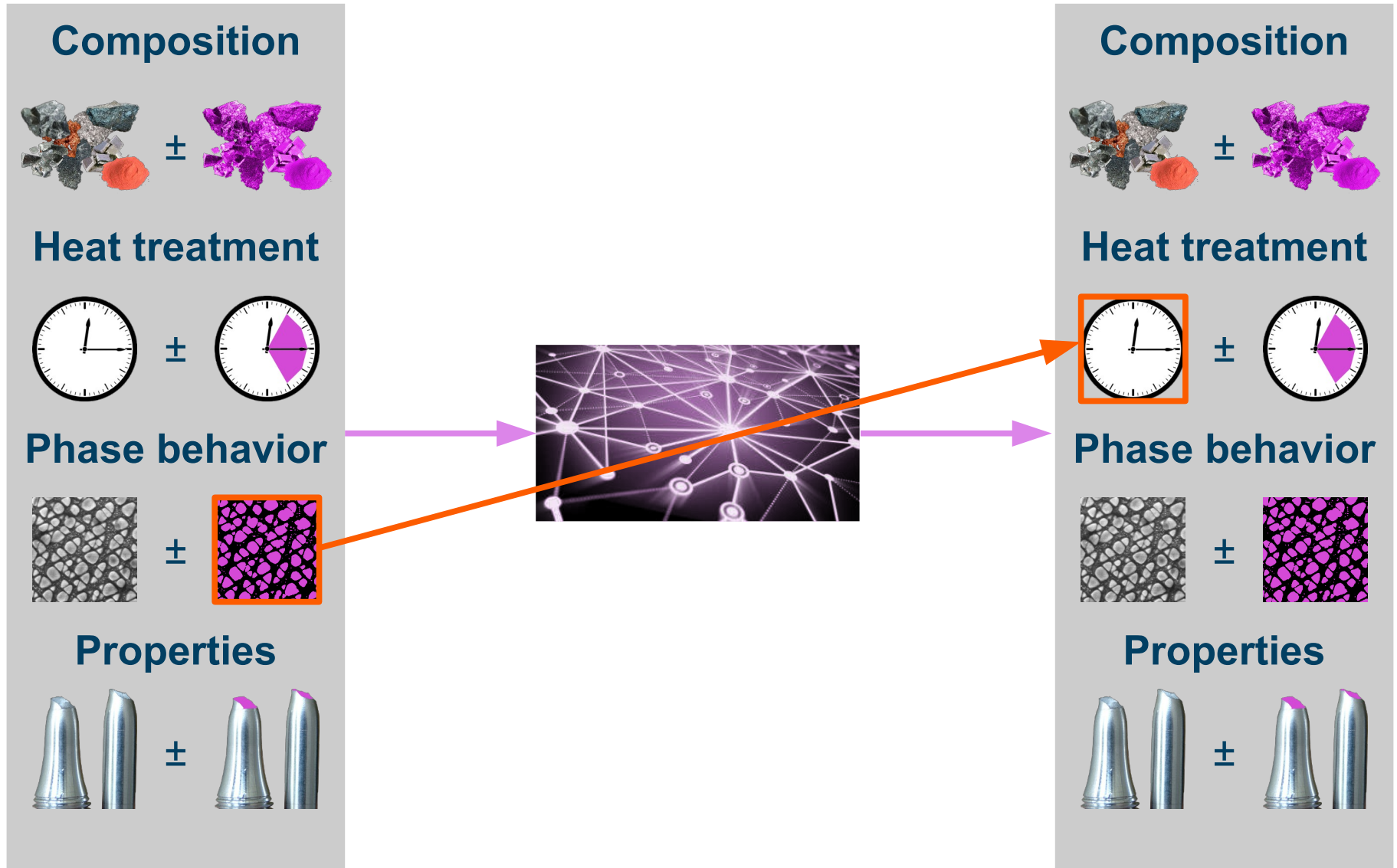
Neural network transmits noise as uncertainty



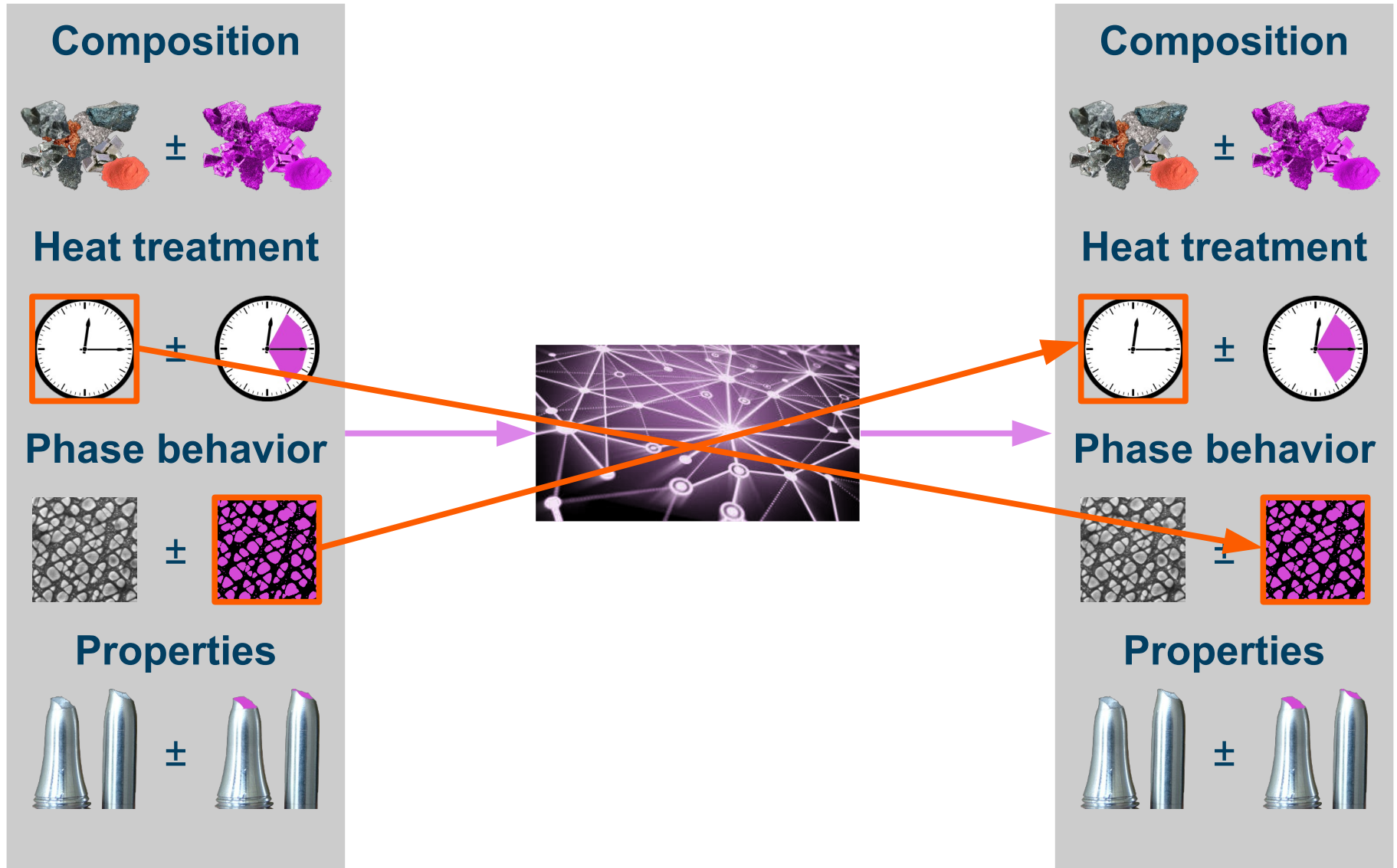
Incorporate noise into the neural network



Exploit noise in the neural network

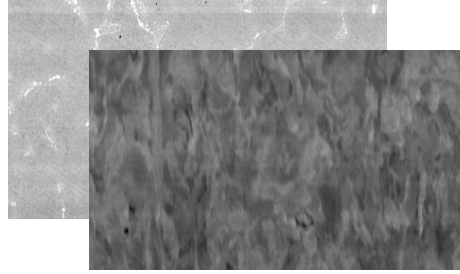


Exploit noise in the neural network

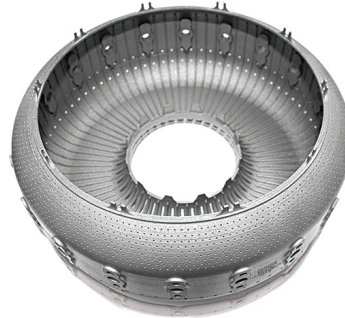


More materials designed: top down

Molybdenum
forging alloys

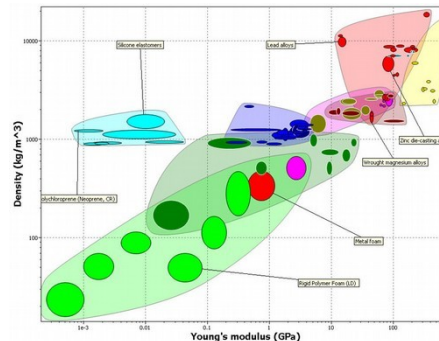


3D printed alloy
designed from
7 data entries



Materials
Solutions

Found 192 errors in
materials databases



GRANTA
MATERIAL INSPIRATION

Even more materials designed: top down and bottom up

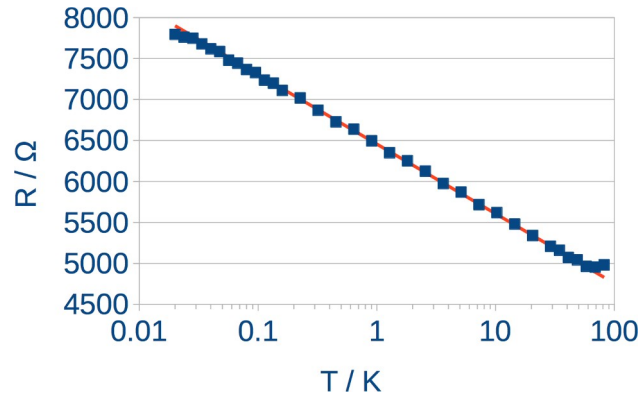
Battery design with DFT and experimental data



Designing lubricants with DFT and experimental data



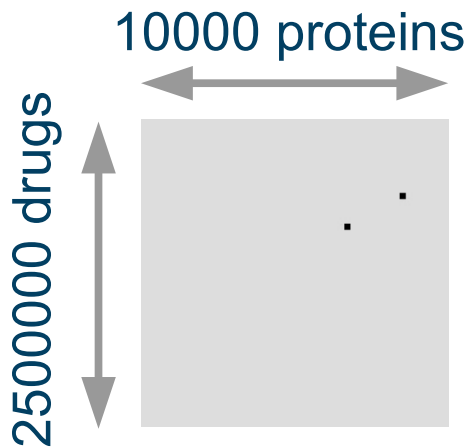
Thermometer with quantum and experimental data



Data available for drug discovery

10,000 proteins with 2,500,000 compounds

Original dataset 0.05% complete



e-therapeutics

Impute the database used for drug discovery

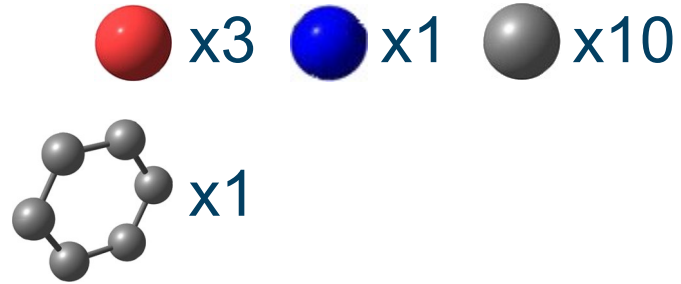
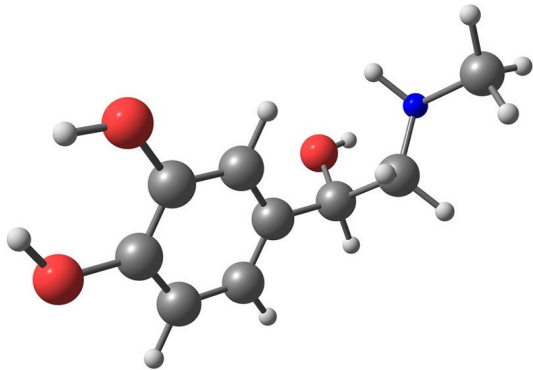
10,000 proteins with 2,500,000 compounds

Original dataset 0.05% complete

Filled 32% of the entries

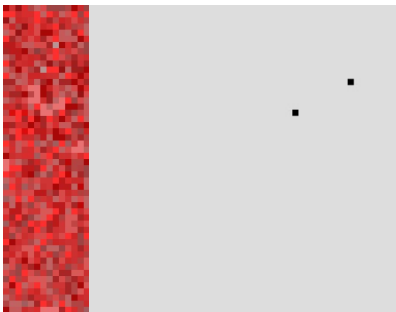


Drug discovery with additional descriptors



1101 Morse descriptor

200



e-therapeutics

Improved drug discovery

Include drug structural information to fill to 46%

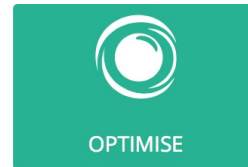
Saved >\$1billion in experimental costs



Startup intellegens productizing the neural network

Input properties - unknowns		
Yield stress / MPa	<input type="text" value="1000.0"/>	Maximize
Ultimate Tensile Strength / MPa	<input type="text" value="1500"/>	Maximize
Elongation	<input type="text" value="10"/>	Minimize

Input composition		
Iron	<input type="text"/>	remain %
Carbon	<input type="text"/>	0 to 0.43 %
Manganese	<input type="text"/>	0 to 3.0 %
Silicon	<input type="text"/>	0 to 4.75 %
Chromium	<input type="text"/>	0 to 17.5 %
Nickel	<input type="text"/>	0 to 21.0 %
Molybdenum	<input type="text"/>	0 to 9.67 %
Vanadium	<input type="text"/>	0 to 4.32 %



Output properties - predicted		
Yield stress	<input type="text" value="1224"/>	± 26 MPa
Ultimate tensile strength	<input type="text" value="1952"/>	± 84 MPa
Elongation	<input type="text" value="7"/>	± 1 %

Output composition		
Iron	<input type="text" value="57.25"/>	%
Carbon	<input type="text" value="0.04"/>	%
Manganese	<input type="text" value="0.02"/>	%
Silicon	<input type="text" value="2.59"/>	%
Chromium	<input type="text" value="11.22"/>	%
Nickel	<input type="text" value="15.05"/>	%
Molybdenum	<input type="text" value="2.45"/>	%
Vanadium	<input type="text" value="0.62"/>	%

Summary: progress

Apply deep learning to high-value **fragmented** data

Exploit knowledge of **noise** in the data

Experimentally **proven** materials design with 7 companies,
founded startup **intellegens**

Summary: opportunities

Apply deep learning to high-value **fragmented** data

Exploit knowledge of **noise** in the data

Experimentally **proven** materials design with 7 companies,
founded startup **intellegens**

Merge experiments and simulations into **holistic** design tool

Summary: challenges

Apply deep learning to high-value **fragmented** data

Exploit knowledge of **noise** in the data

Experimentally **proven** materials design with 7 companies,
founded startup **intellegens**

Merge experiments and simulations into **holistic** design tool

Scientists establish all possible **SOURCES** of information