



Materials discovery with artificial intelligence

Gareth Conduit

Neural network algorithm to

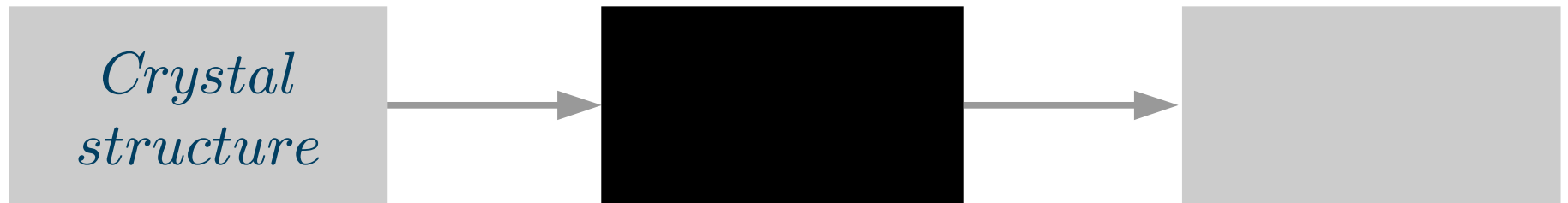
Merge simulations, physical laws, and experimental data

Reduce the need for expensive experimental development

Accelerate materials and drugs discovery

Generic with **proven** applications in materials discovery and drug design

Neural network: train on complete data



Neural network: train on complete data



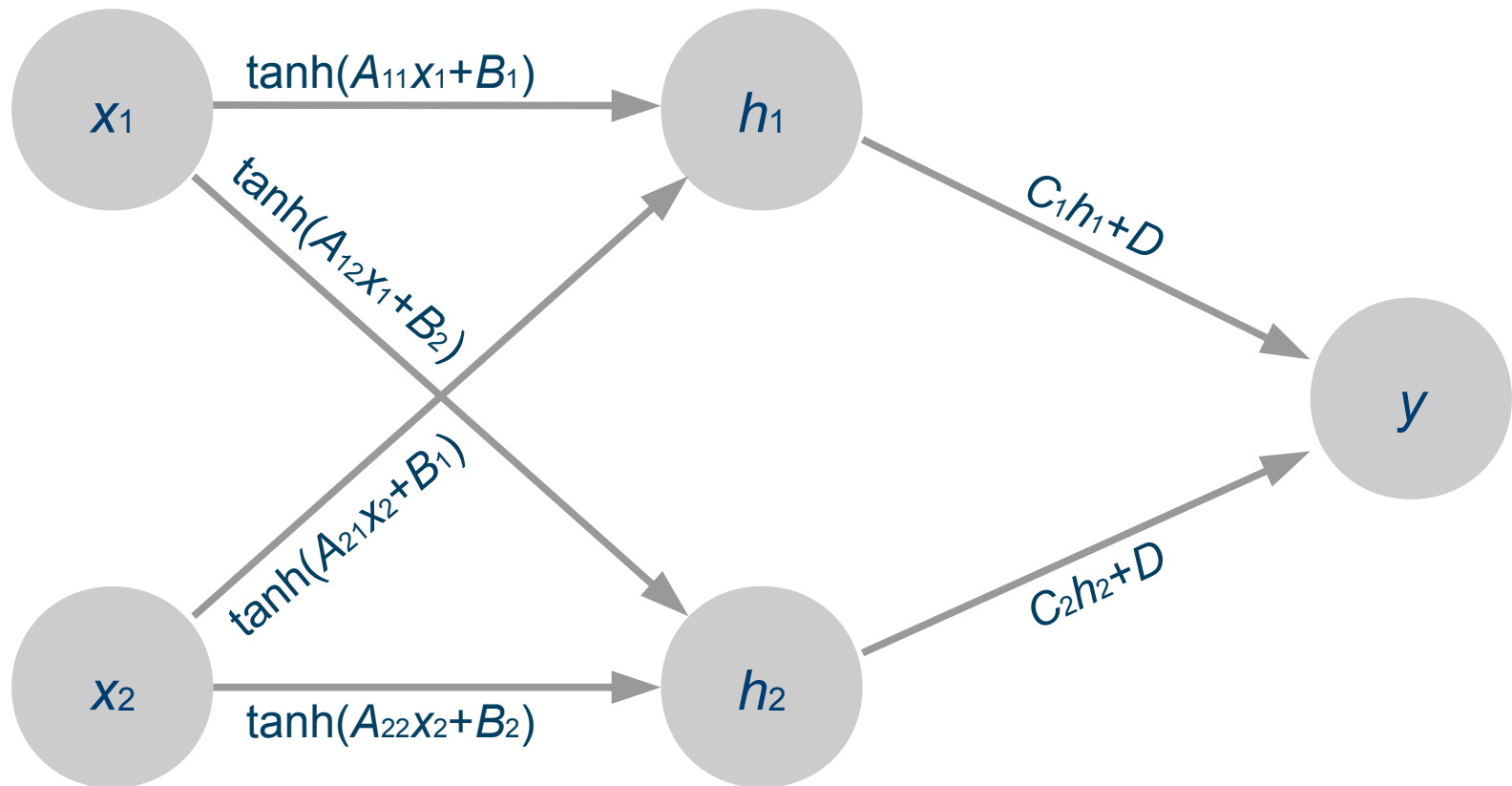
Neural network: train on complete data

*Crystal
structure*



Crystal
structure

Neural networks: mathematical form



Proposed neural network: train on fragmented data

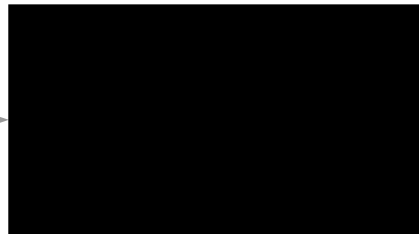
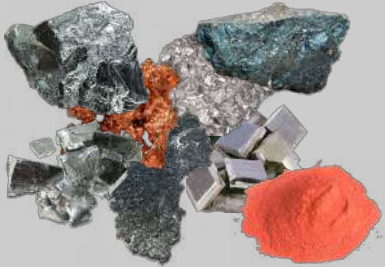


Proposed neural network: predict on fragmented data



Black box for materials design

Composition



Properties


UTS

Hardness

Cost

Training the neural network

Composition



Properties

UTS

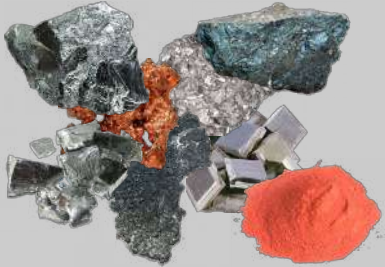
Hardness

Cost

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488685276110993
203332721994995
976579342243418
394046703960393
597692868112392
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805556069526643
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Neural network for materials design

Composition



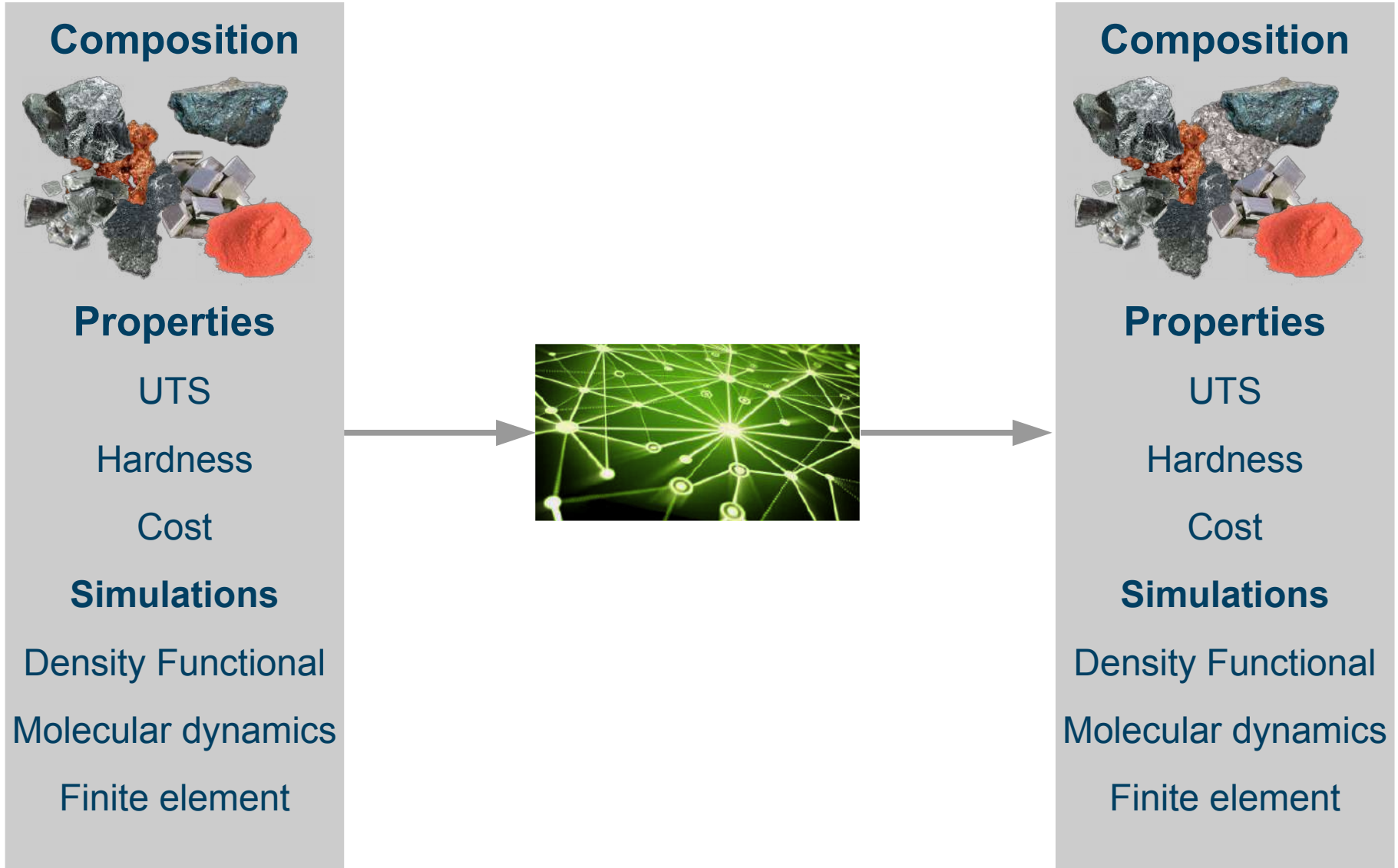
Properties

UTS

Hardness

Cost

Neural network to exploit all available correlations



Fragmented training data set

Composition

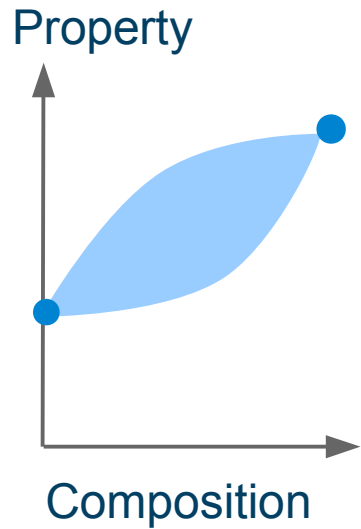
UTS

Hardness



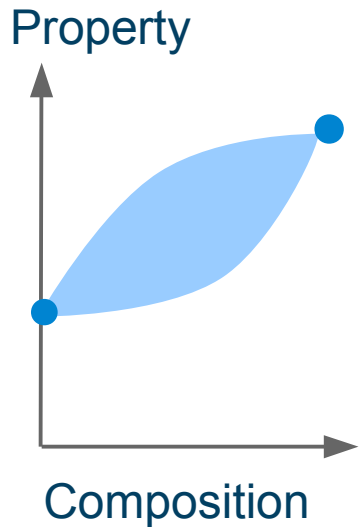
Neural network trained on experimental data

Experiment

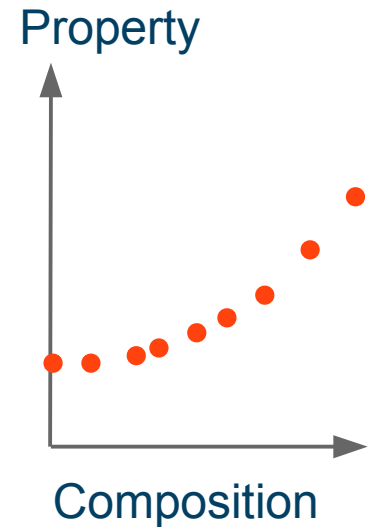


Further information is provided by a simulation

Experiment

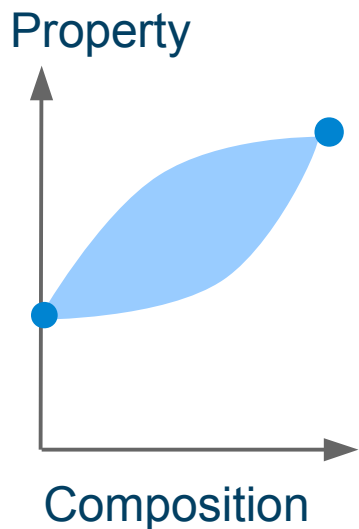


Simulation

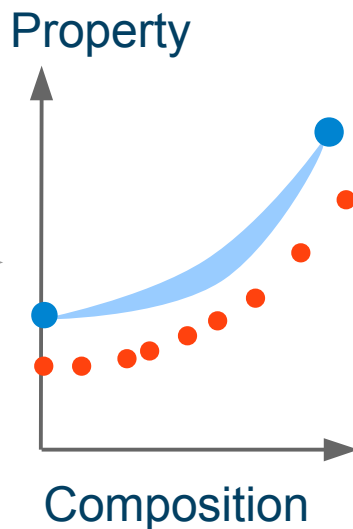


Neural network combines the two sources of data

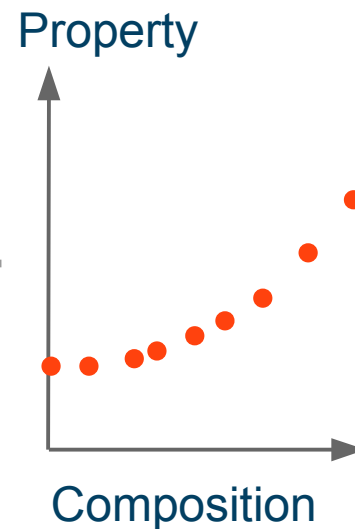
Experiment



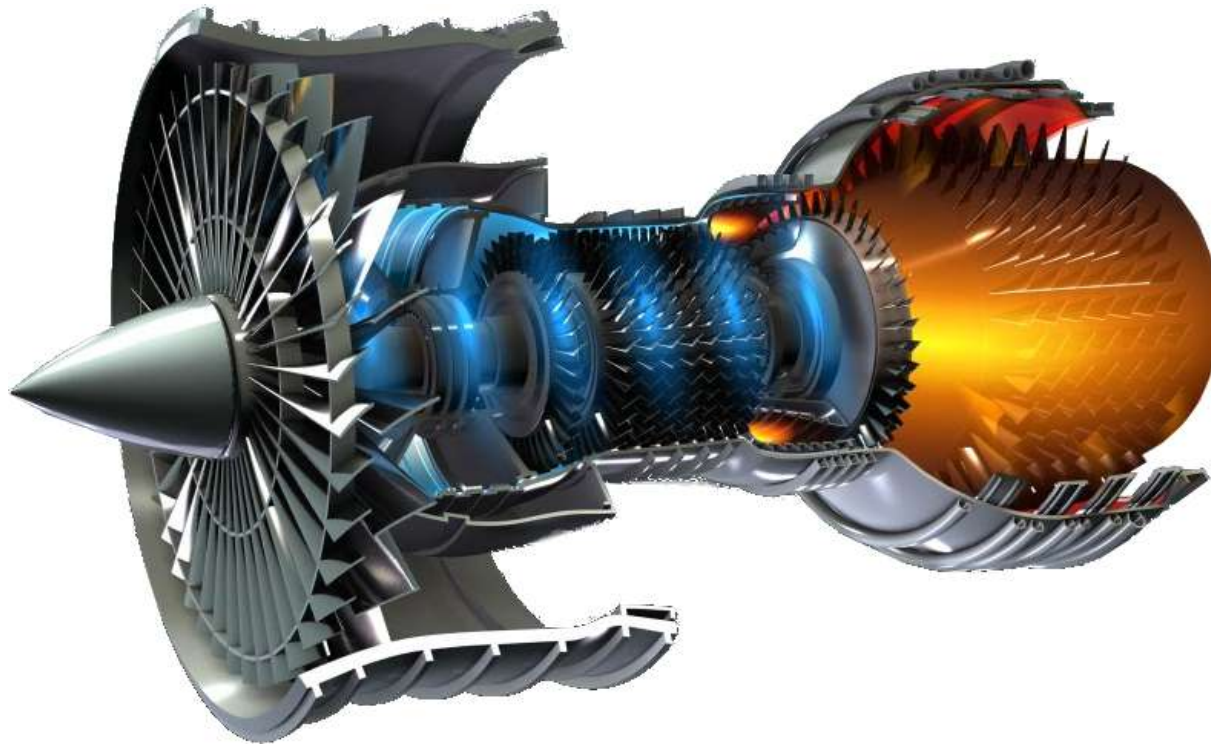
Combined



Simulation



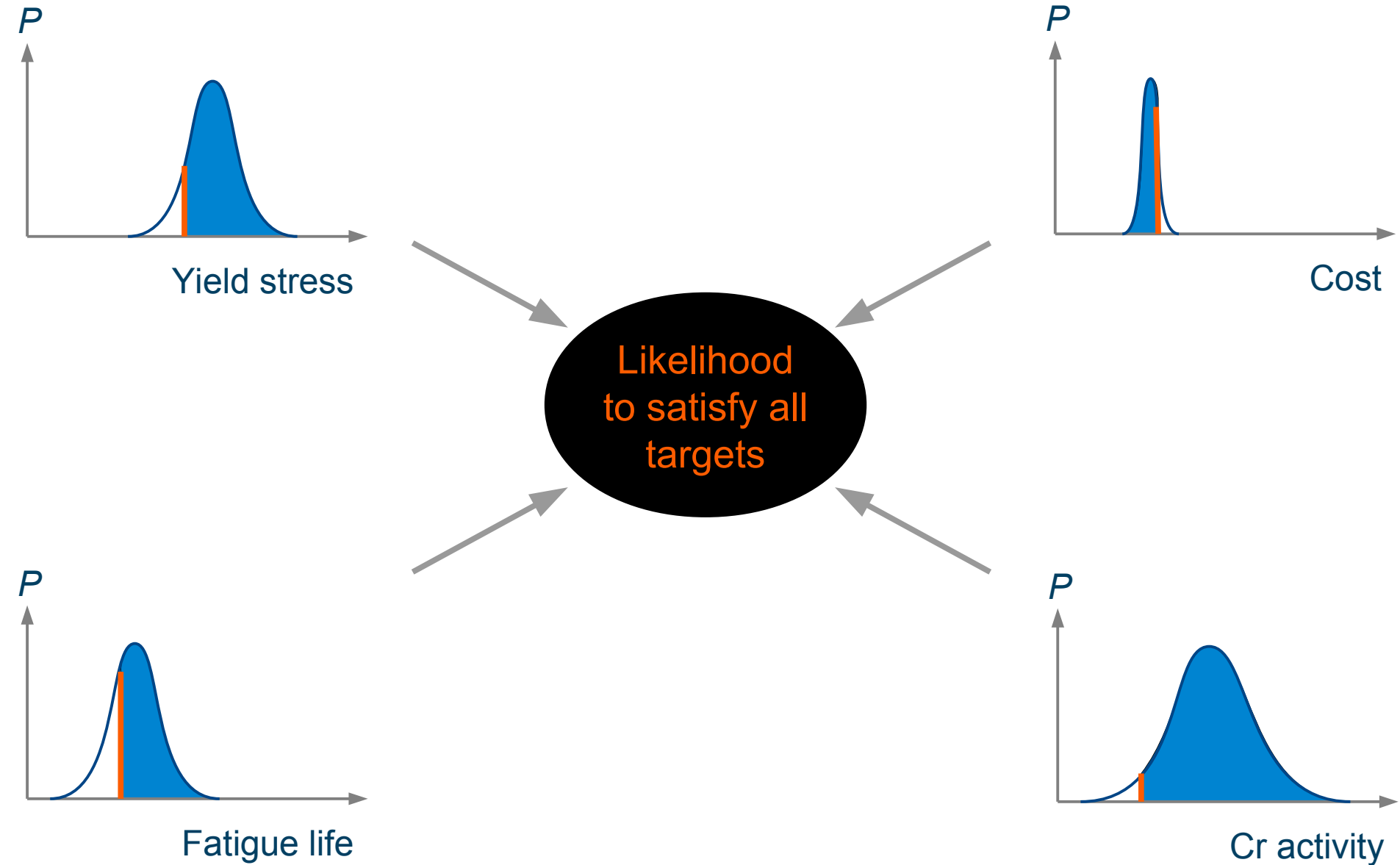
Schematic of an engine



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



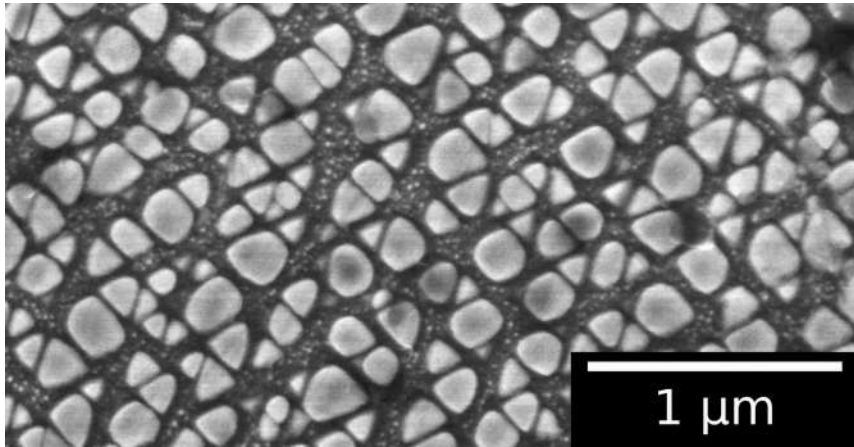
900°C



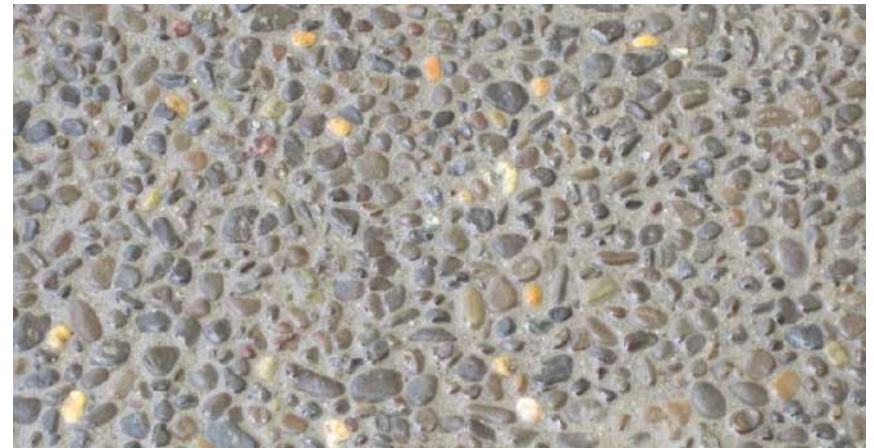
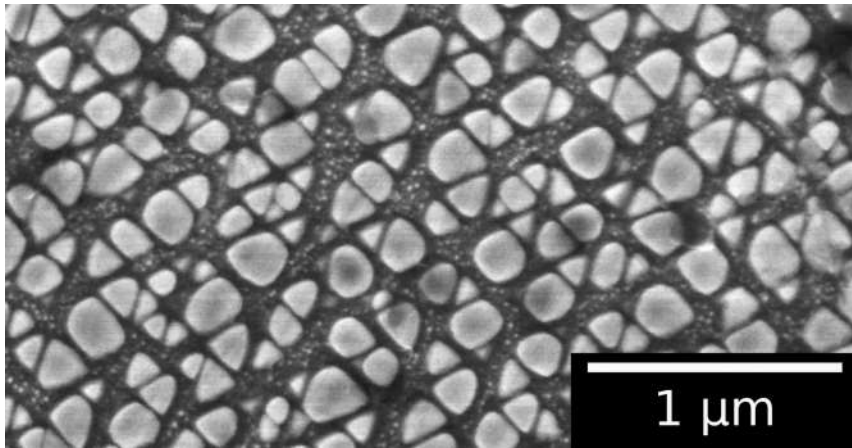
30 hours



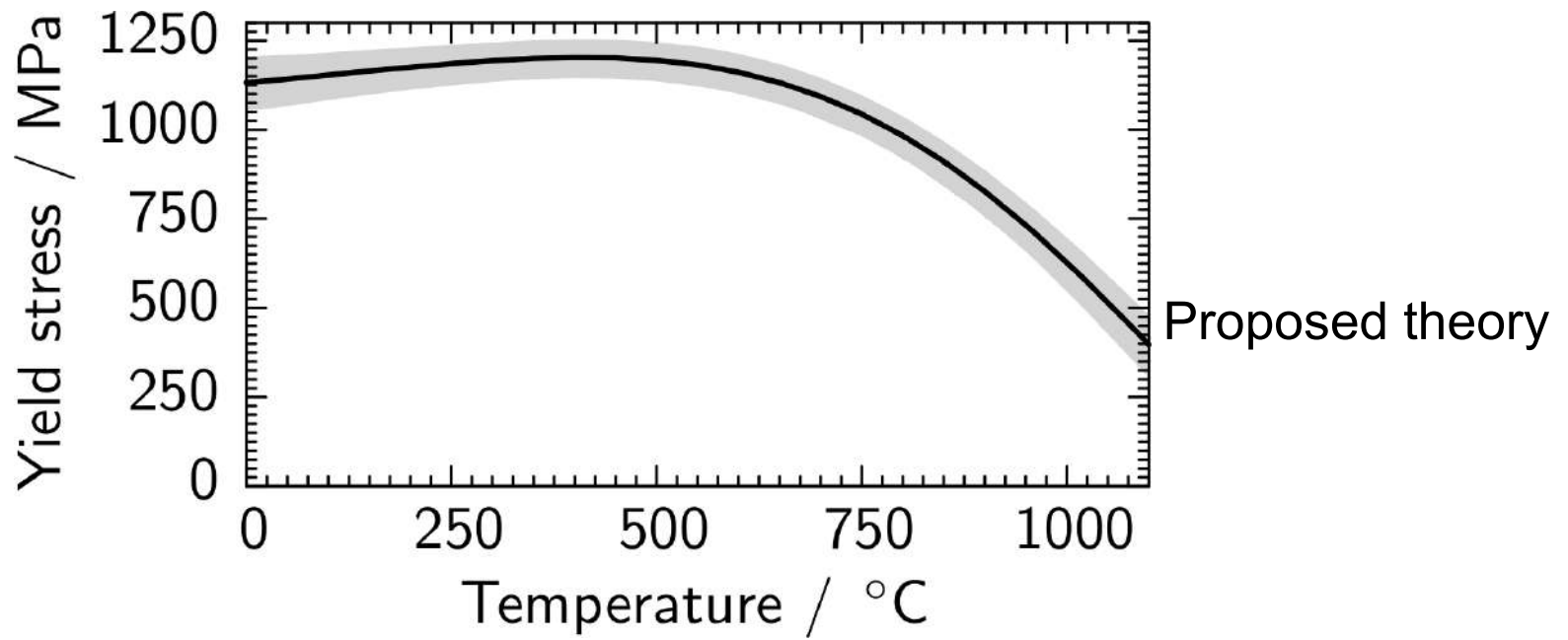
Microstructure



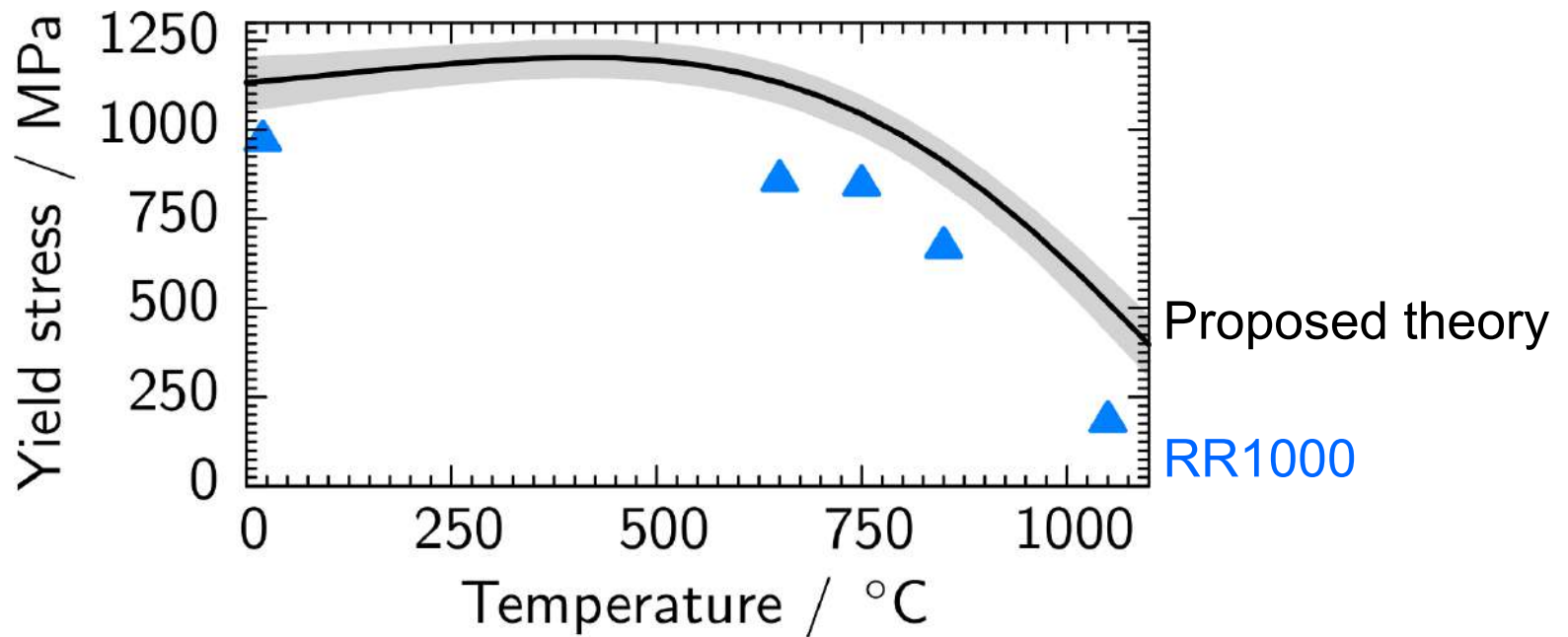
Precipitates strengthen the alloy



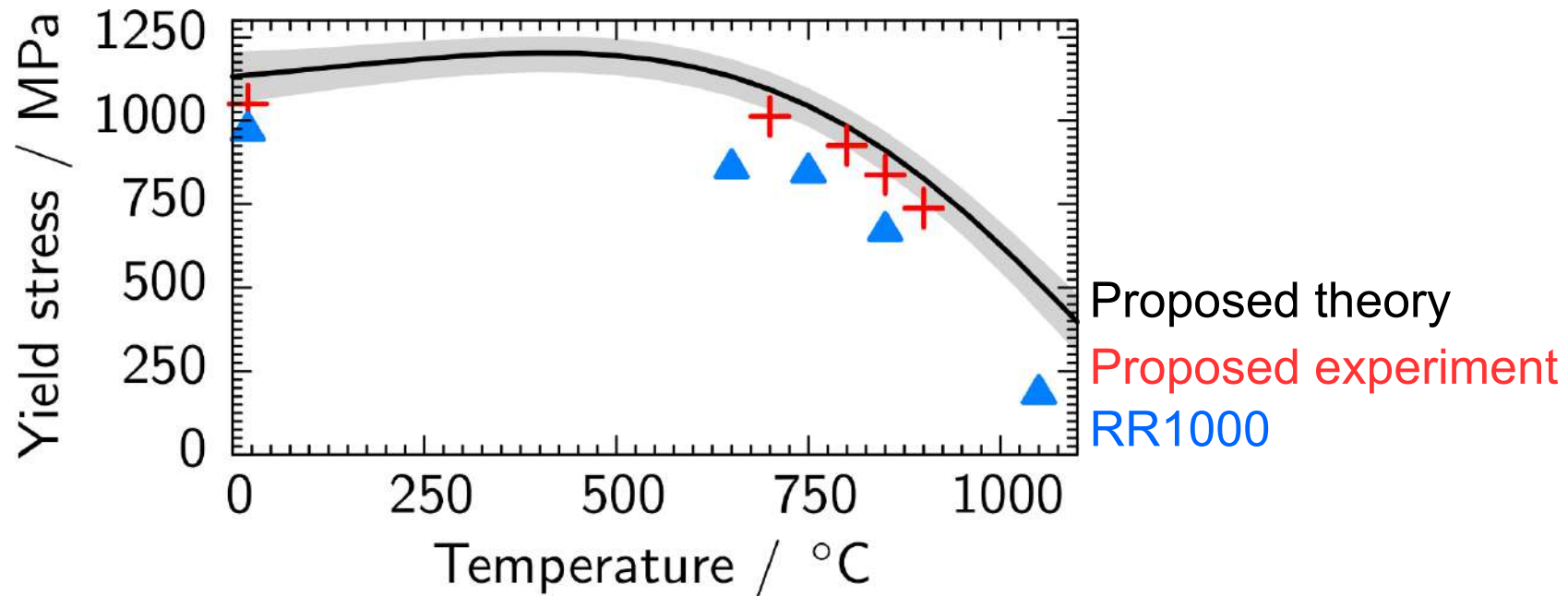
Predict the yield stress



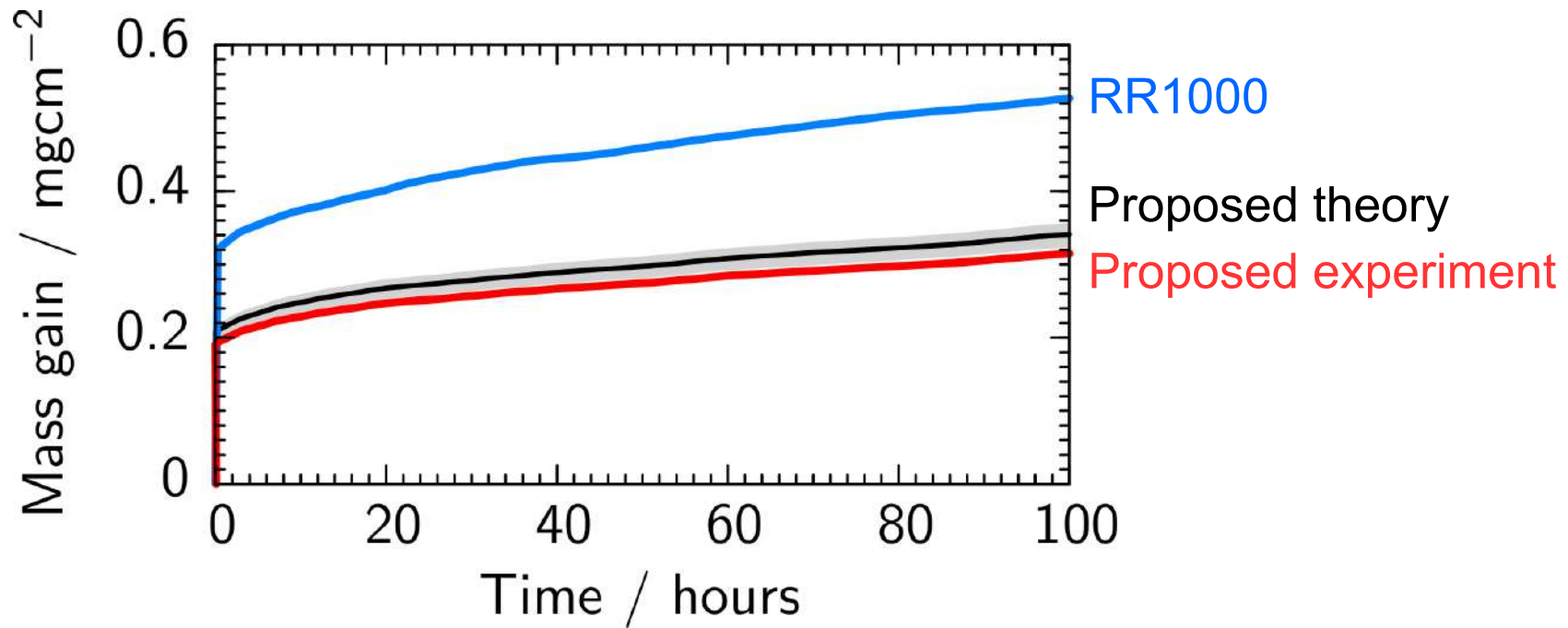
Test the yield stress



Test the yield stress

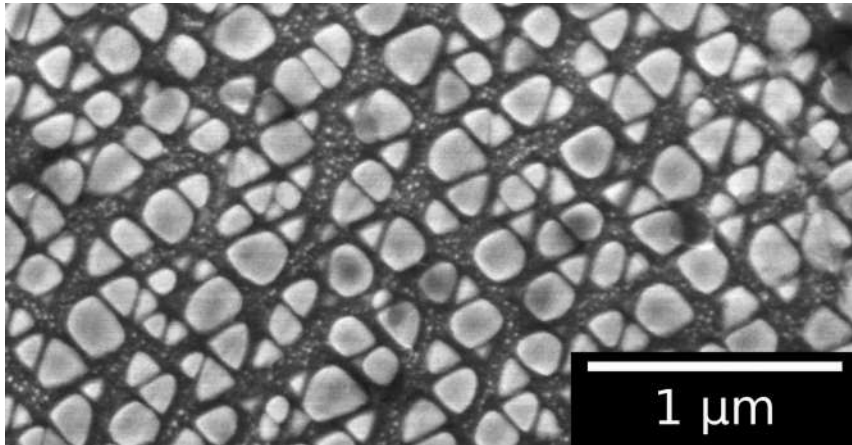


Test the oxidation resistance

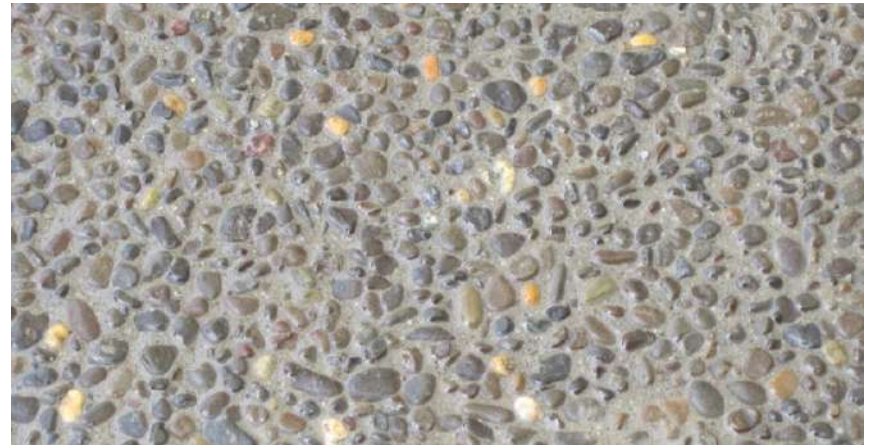


Microstructure strengthens the alloy

Precipitates

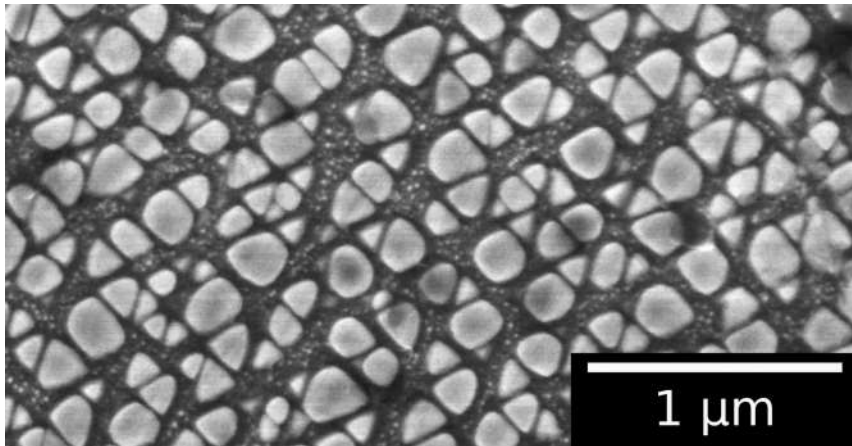


Aggregate

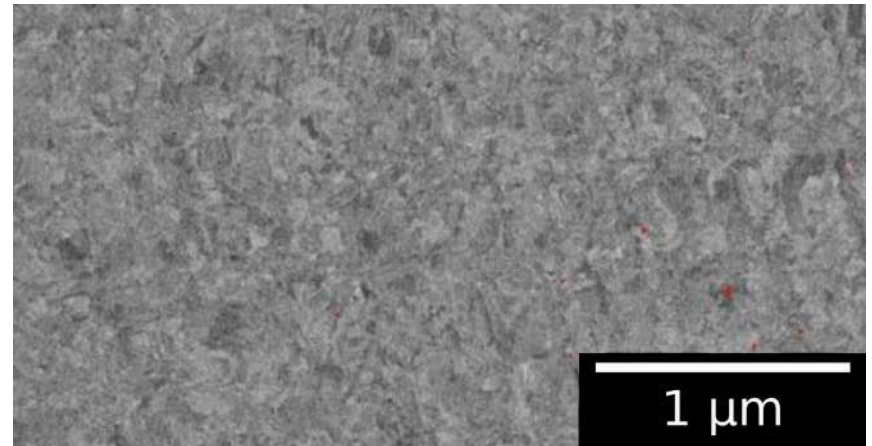


Microstructure defined by the heat treatment

Correct heat treatment

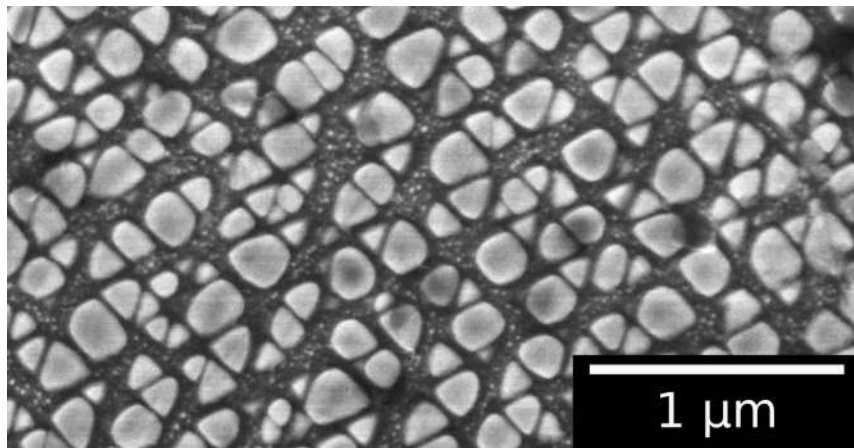


Incorrect heat treatment

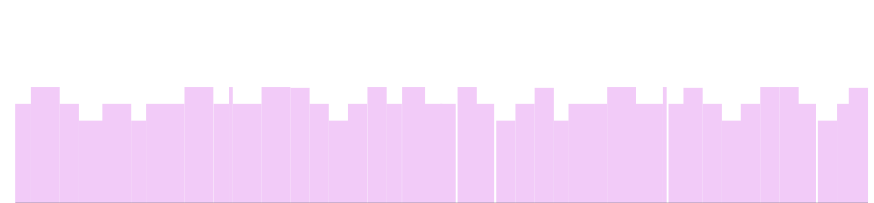
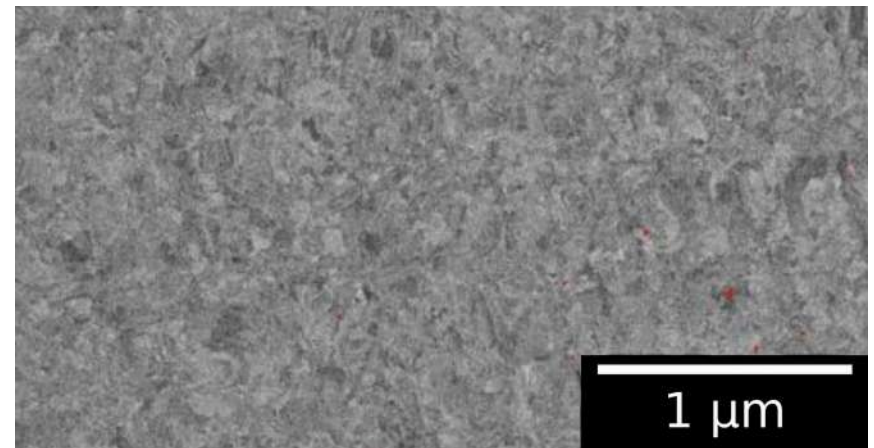


Microstructure distribution links to the heat treatment

Correct heat treatment

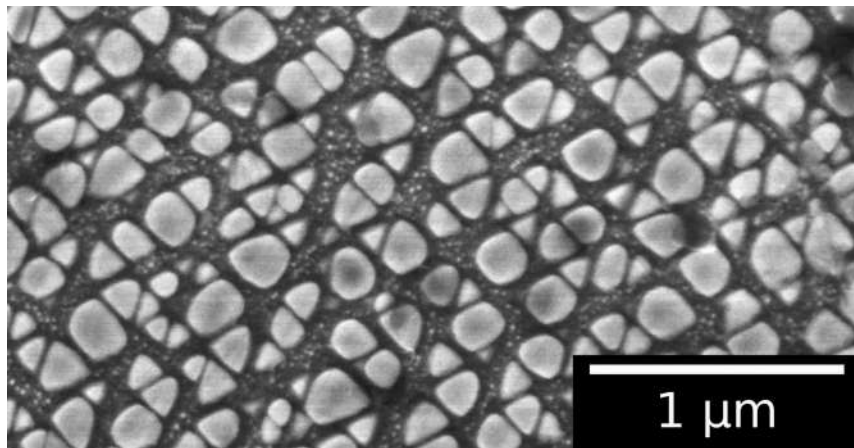


Incorrect heat treatment

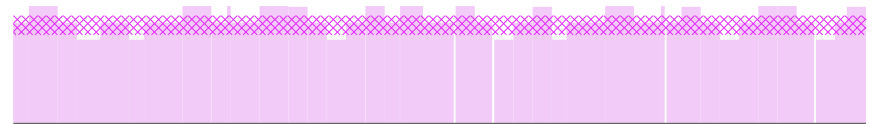
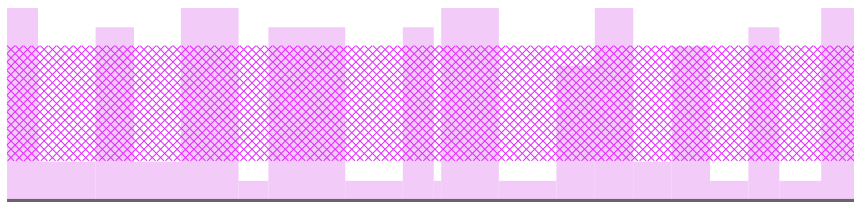
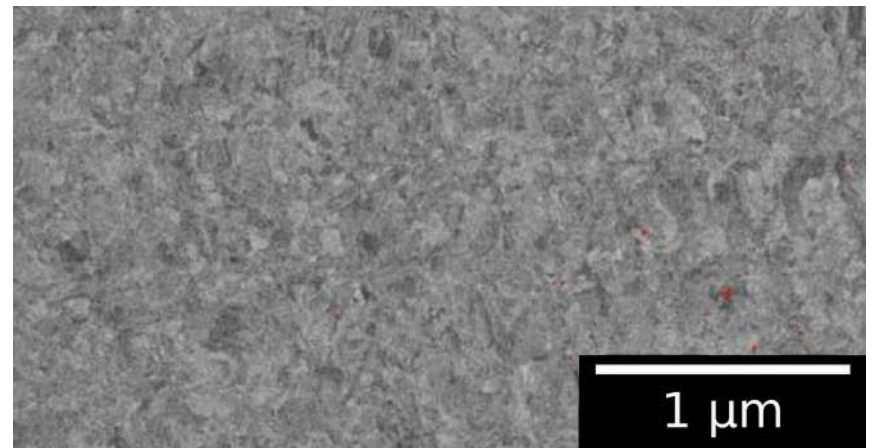


Microstructure distribution links to the heat treatment

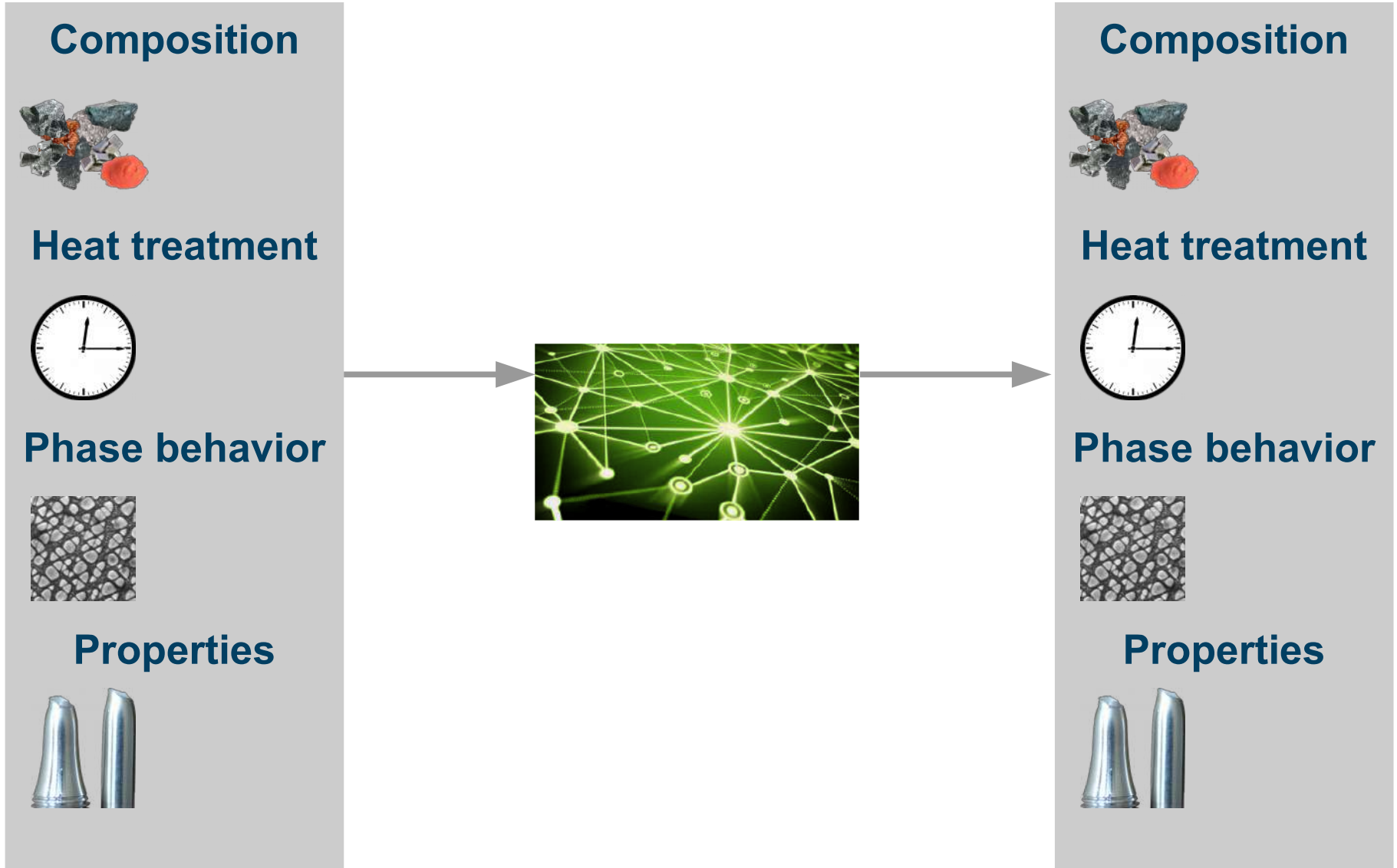
Correct heat treatment



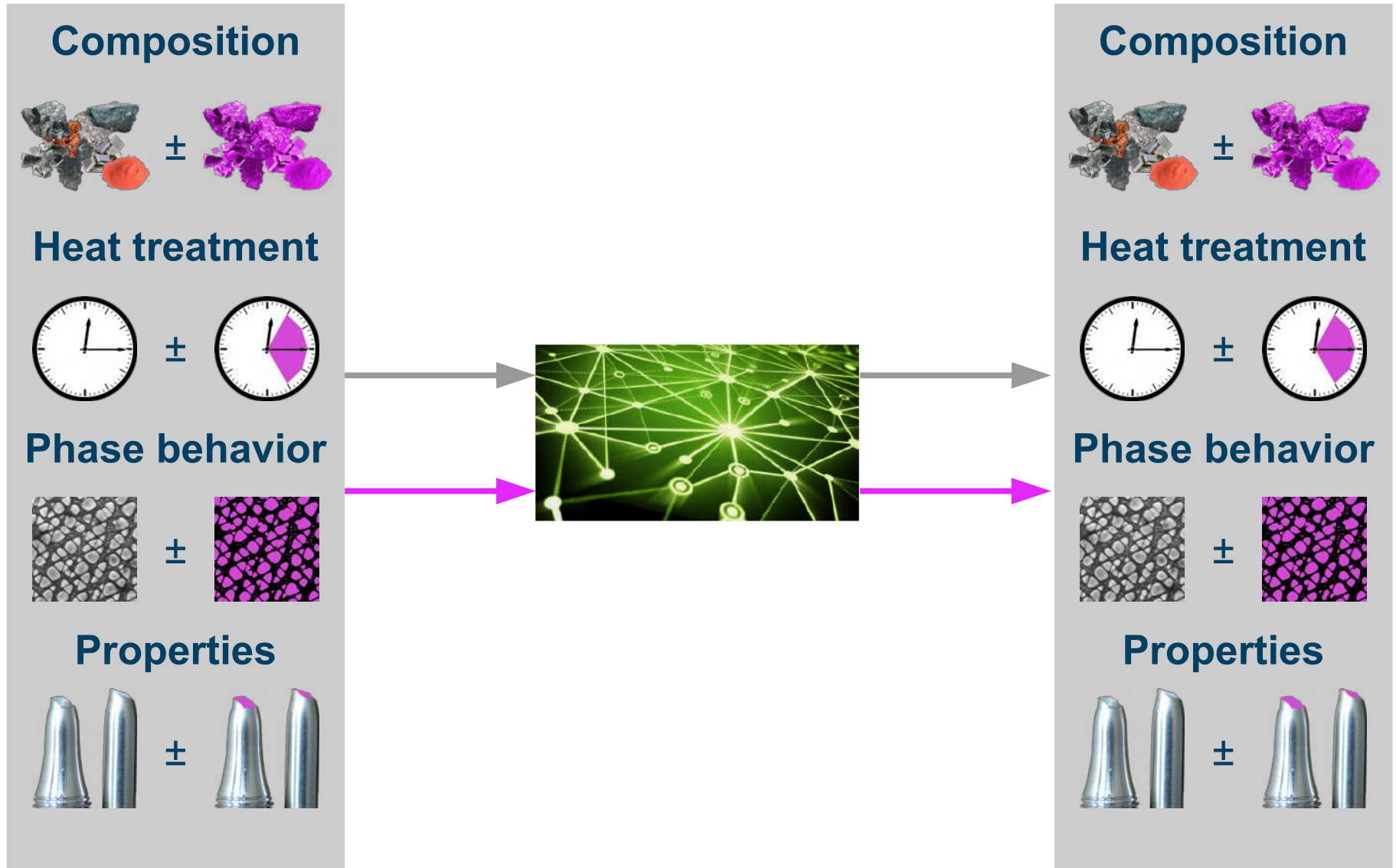
Incorrect heat treatment



Standard neural network

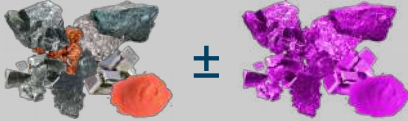


Neural network transmits noise as uncertainty

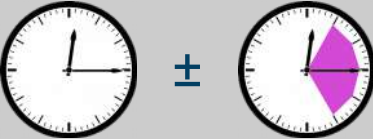


Incorporate noise into the neural network

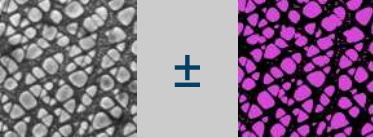
Composition




Heat treatment




Phase behavior




Properties



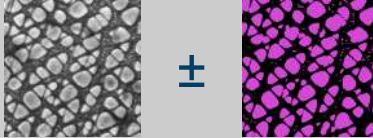
Composition




Heat treatment



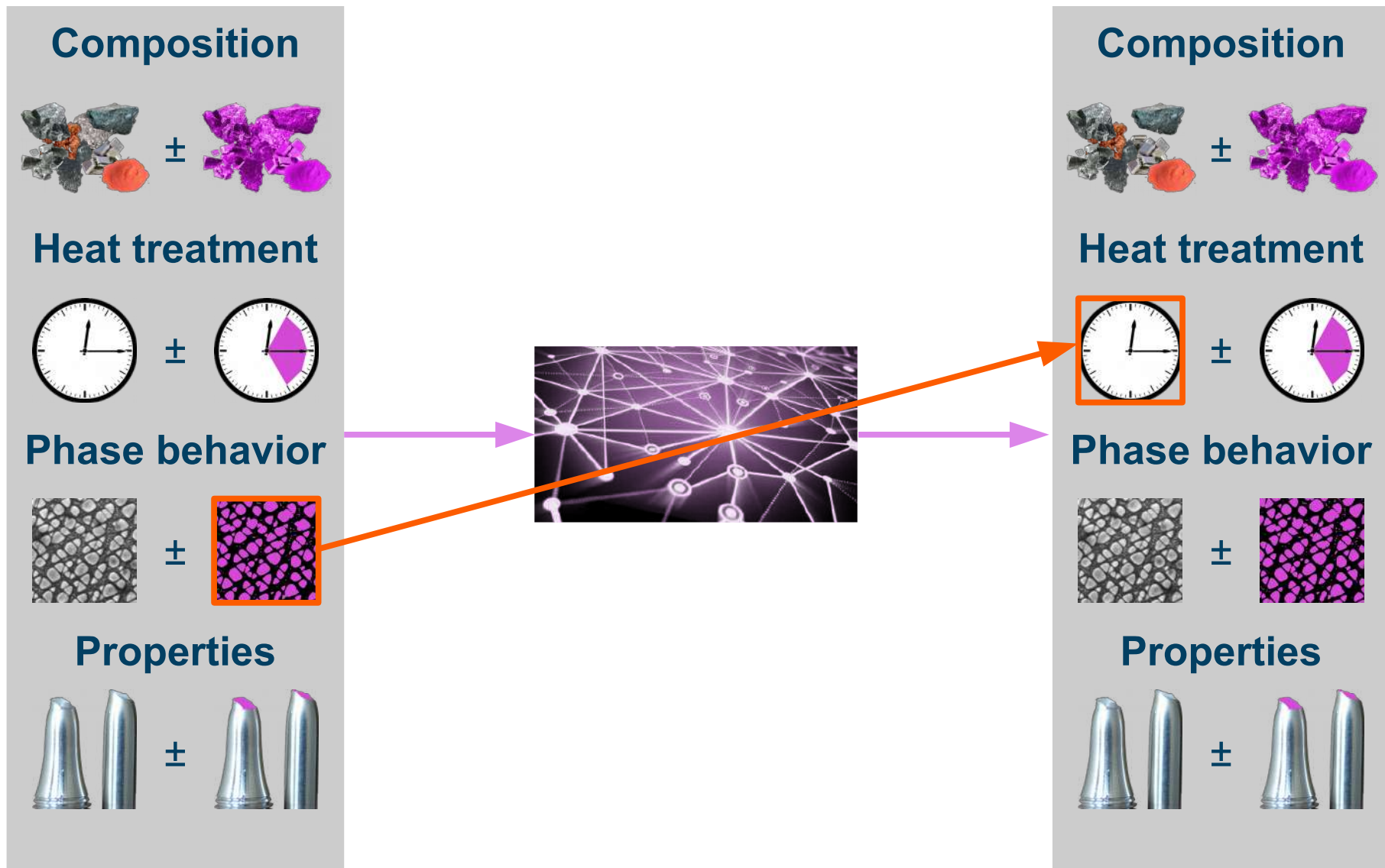
Phase behavior



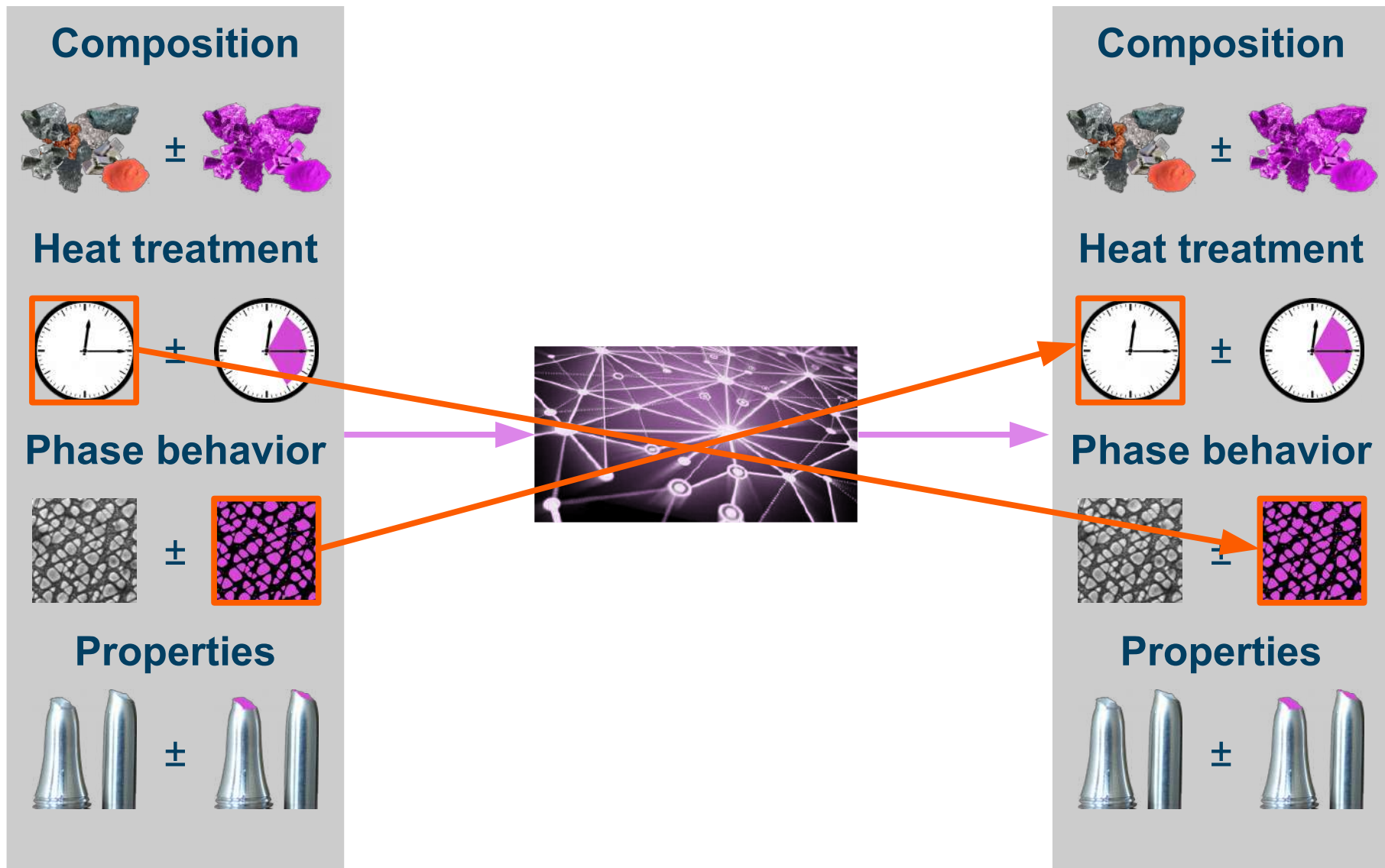
Properties



Exploit noise in the neural network



Exploit noise in the neural network



Point cloud: noise in the data

Near

Far



Point cloud: benefits of including noise

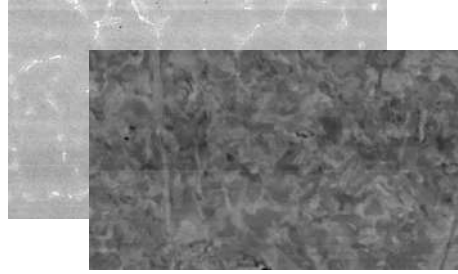
Recover presence of people to **90%** accuracy

Also applies to **trees** and **railings**

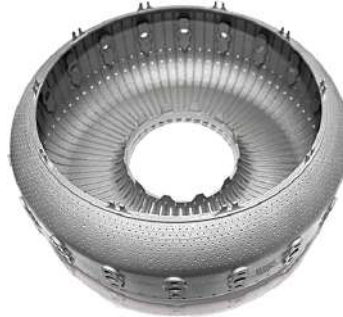


More materials designed

Molybdenum
forging alloys

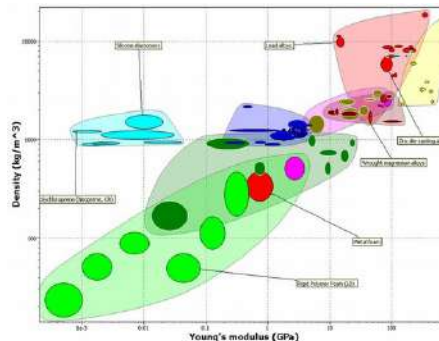


3D printed alloy
designed from
7 data entries



 **Materials
Solutions**

Found 192 errors in
materials databases



GRANTA
MATERIAL INSPIRATION

Even more materials designed

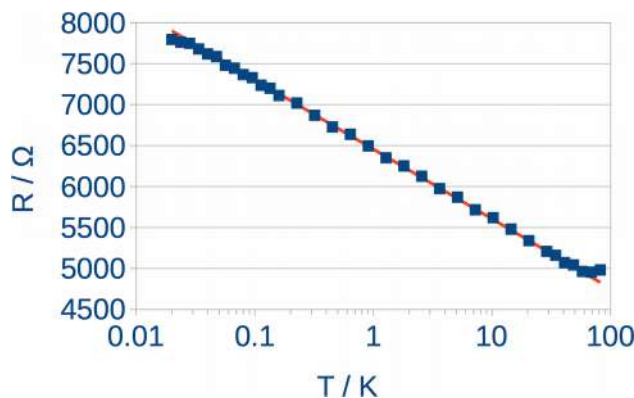
Battery design
with DFT and
experimental data



Designing lubricants
with DFT and
experimental data



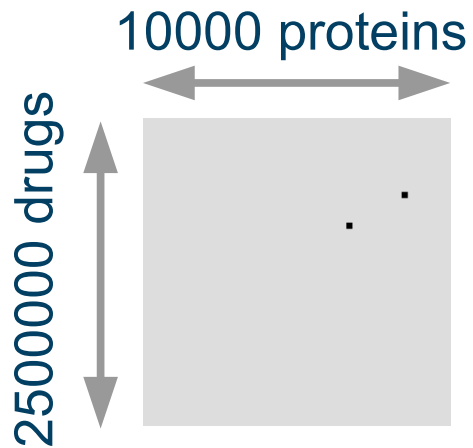
Thermometer with
quantum and
experimental data



Data available for drug discovery

10,000 proteins with 2,500,000 compounds

Original dataset 0.05% complete



Impute the database used for drug discovery

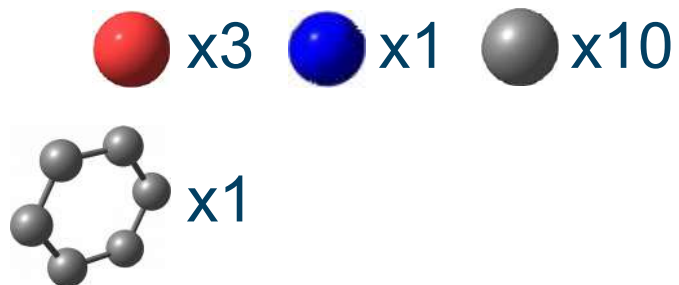
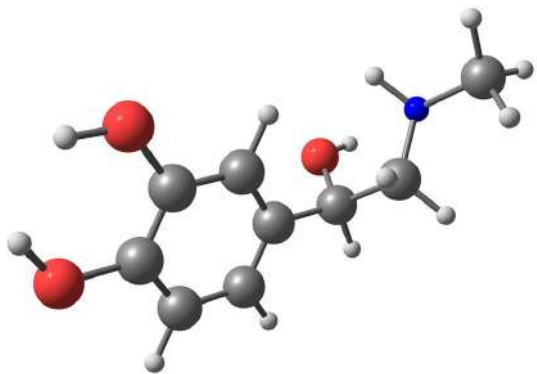
10,000 proteins with 2,500,000 compounds

Original dataset 0.05% complete

Filled 32% of the entries

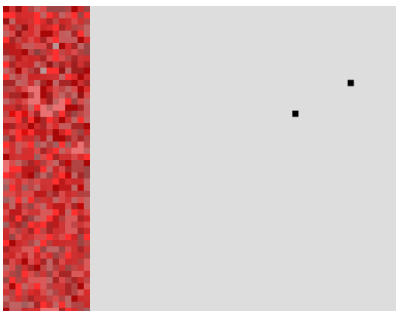


Drug discovery with additional descriptors



1101 Morse descriptor

200



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"/>	<input type="button" value="Maximize"/>
Ultimate Tensile Strength / MPa	<input type="text" value="1500"/>	<input type="button" value="Maximize"/>
Elongation	<input type="text" value="10"/>	<input type="button" value="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"/>	<input type="button" value="± 26"/>	MPa
Ultimate tensile strength	<input type="text" value="1952"/>	<input type="button" value="± 84"/>	MPa
Elongation	<input type="text" value="7"/>	<input type="button" value="± 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 **probability distribution** of the data

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

Summary: opportunities

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

Exploit knowledge of **probability distribution** of the data

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

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

Summary: future prospects

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

Exploit knowledge of **probability distribution** of the data

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

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

Engineers establish all possible **sources** of information