

Concurrent materials design

Gareth Conduit

EP14153898.3; US 2014/177578; GB1302743.8

EP14161255.6; US 2014/223465; GB1307533.8

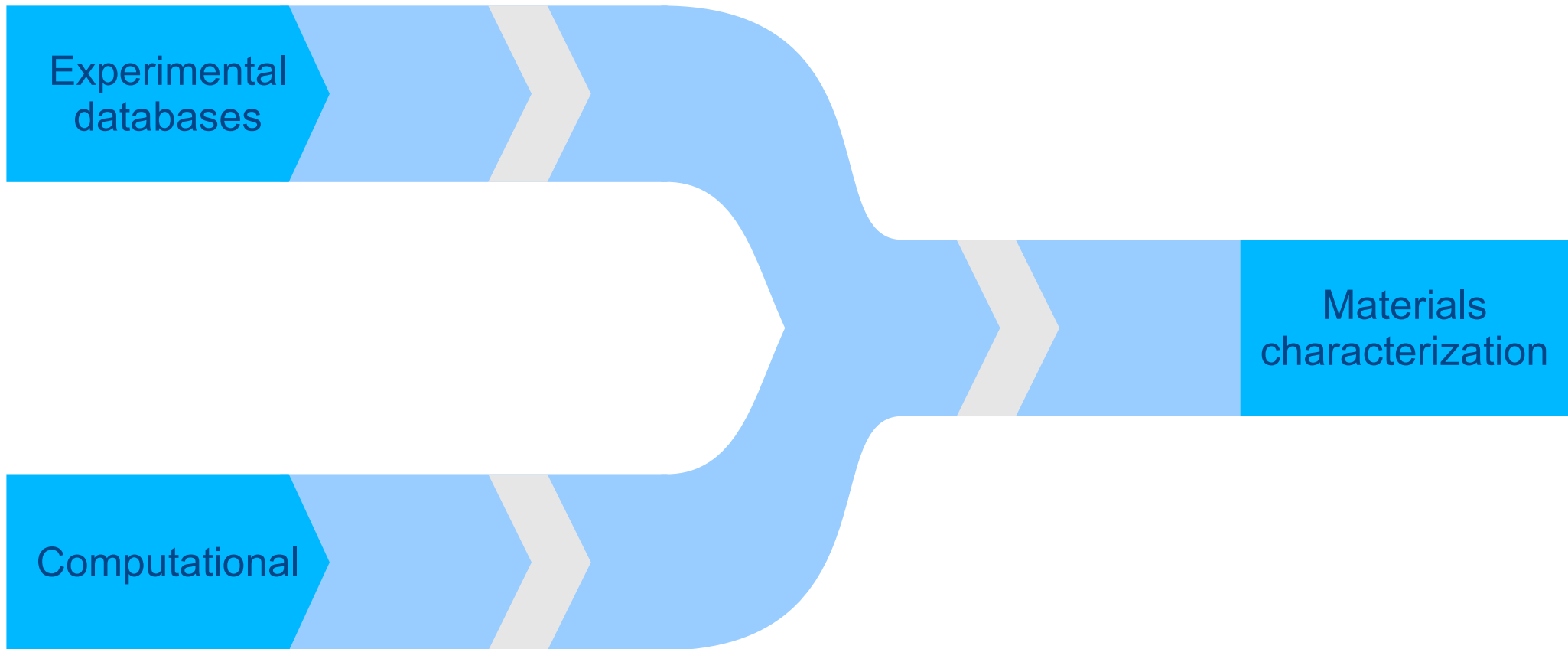
EP14161529.4; GB1307535.3

EP14157622.3; amendment to US 2013/0052077 A1; GB1408536.9

Acta Materialia **61**, 3378 (2013)

Intermetallics **48**, 62 (2014)

Materials pipeline

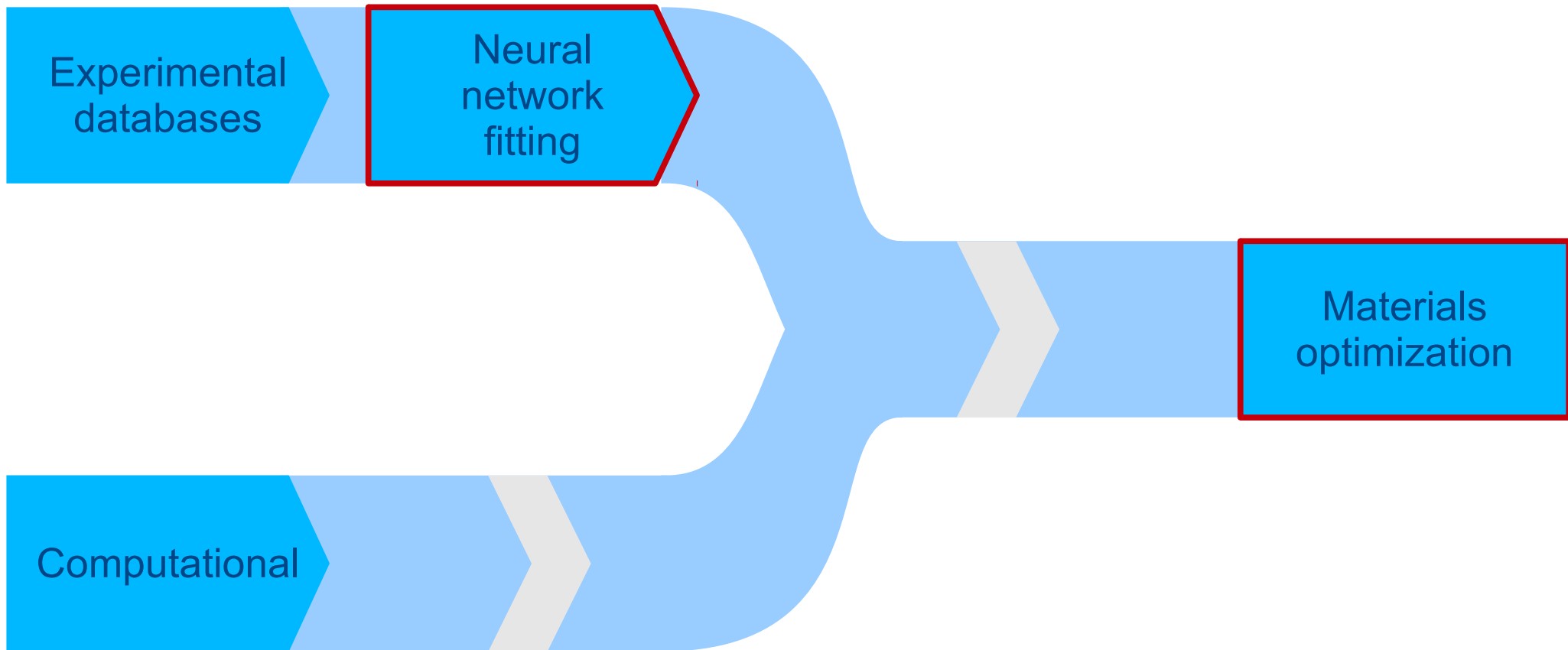


Experimental
databases

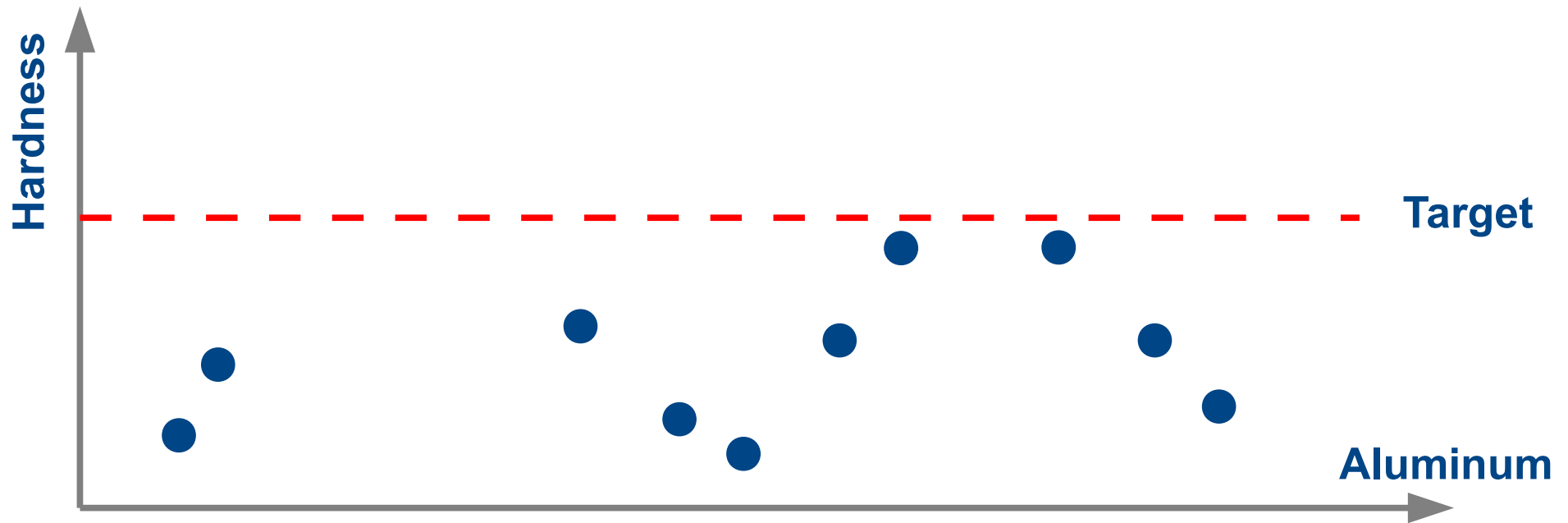
Computational

Materials
characterization

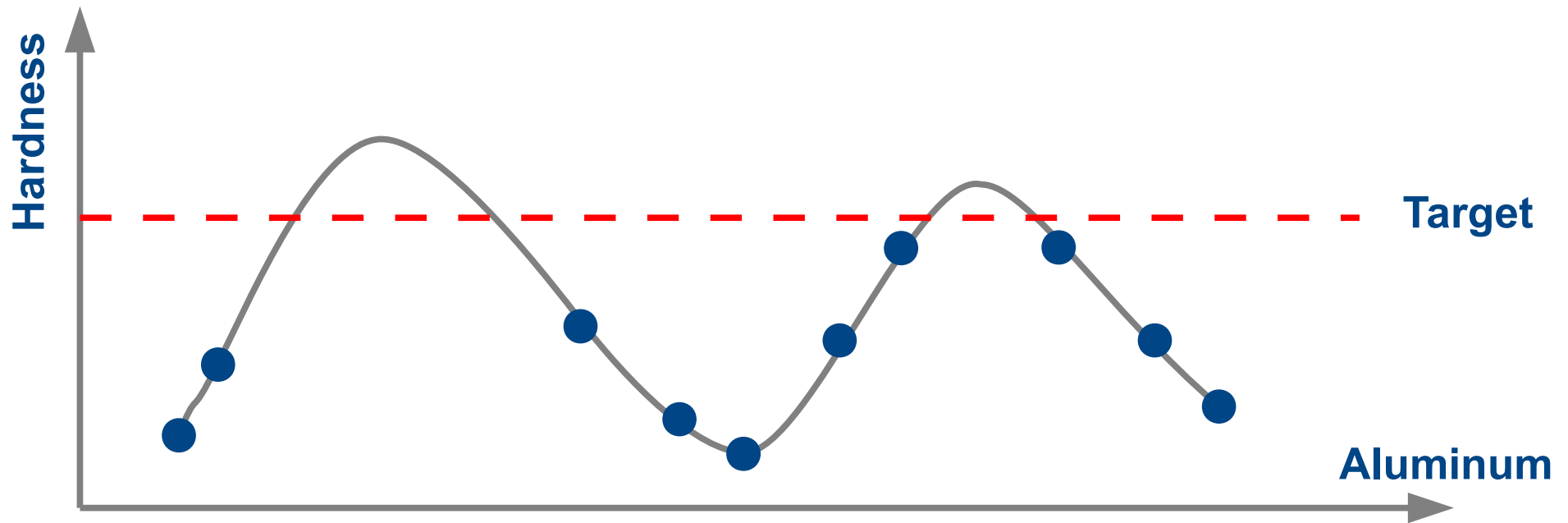
Two new tools



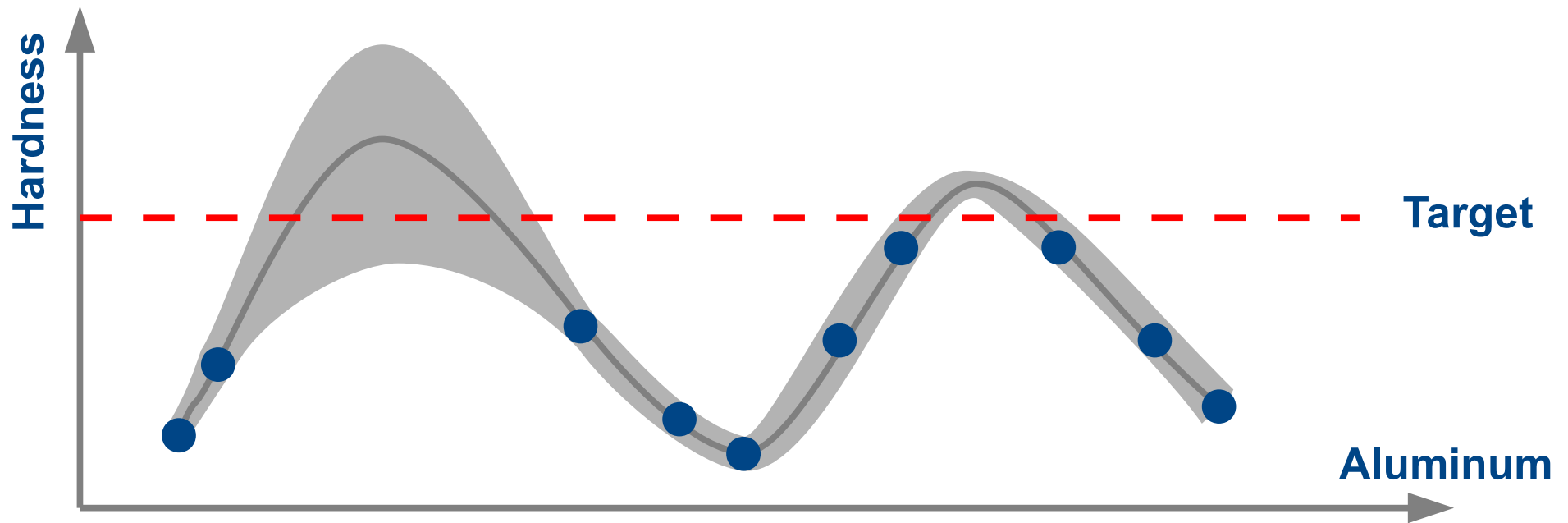
Neural network fitting & optimization



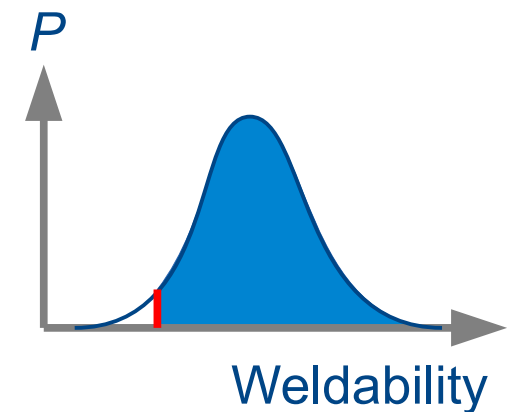
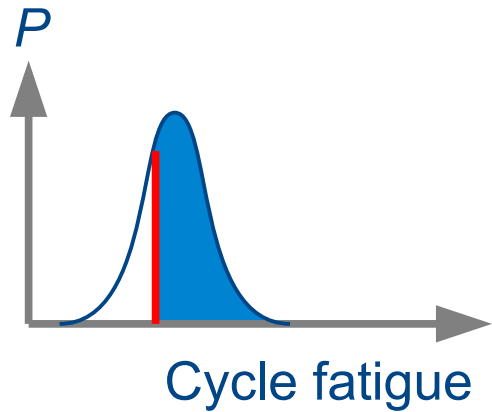
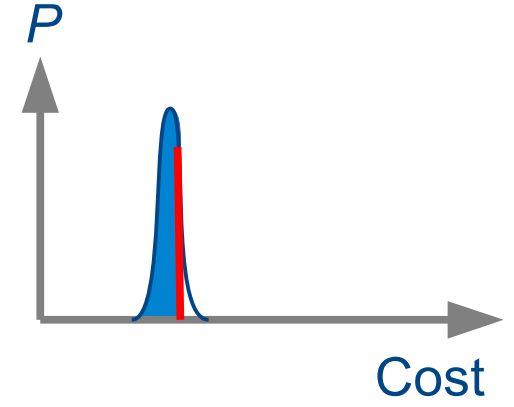
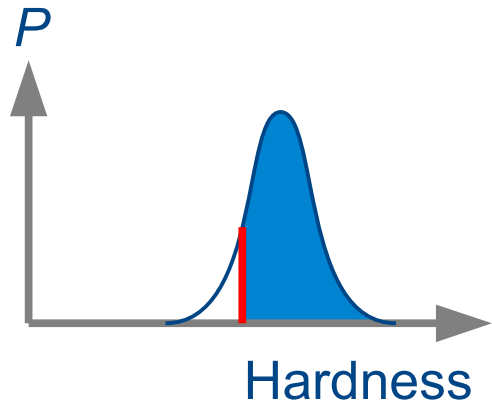
Neural network fitting & optimization



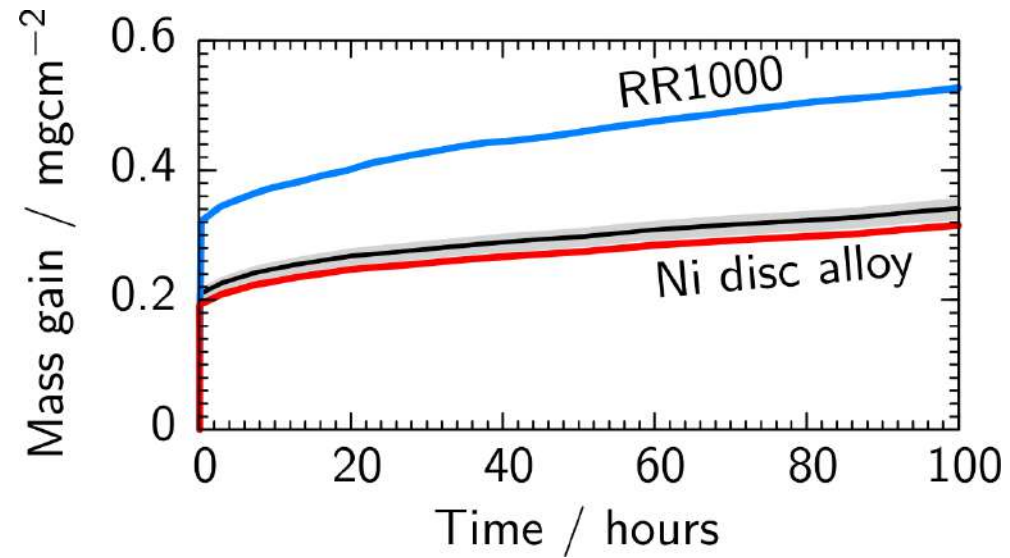
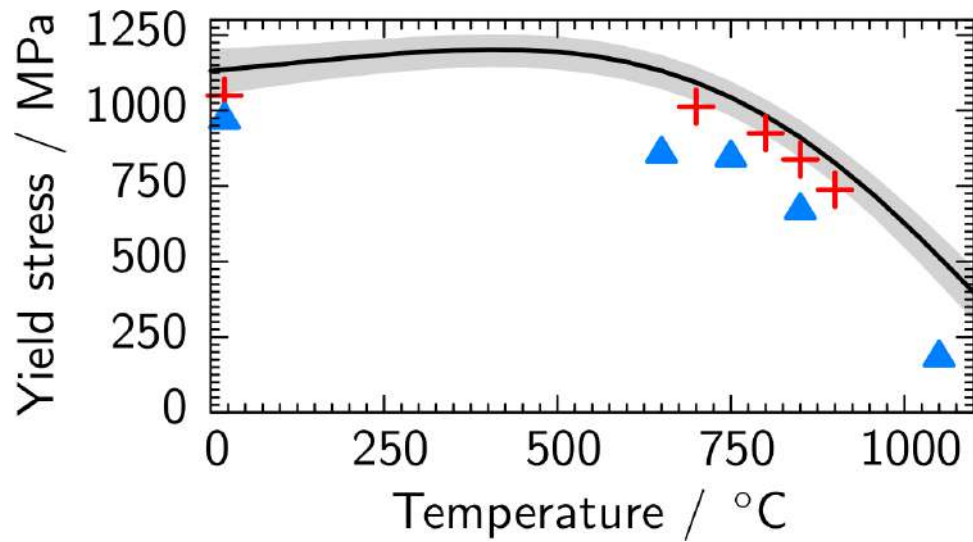
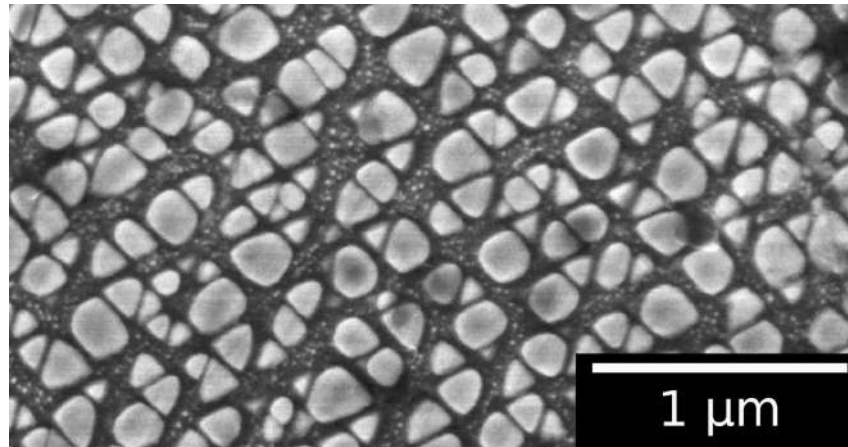
Neural network fitting & optimization



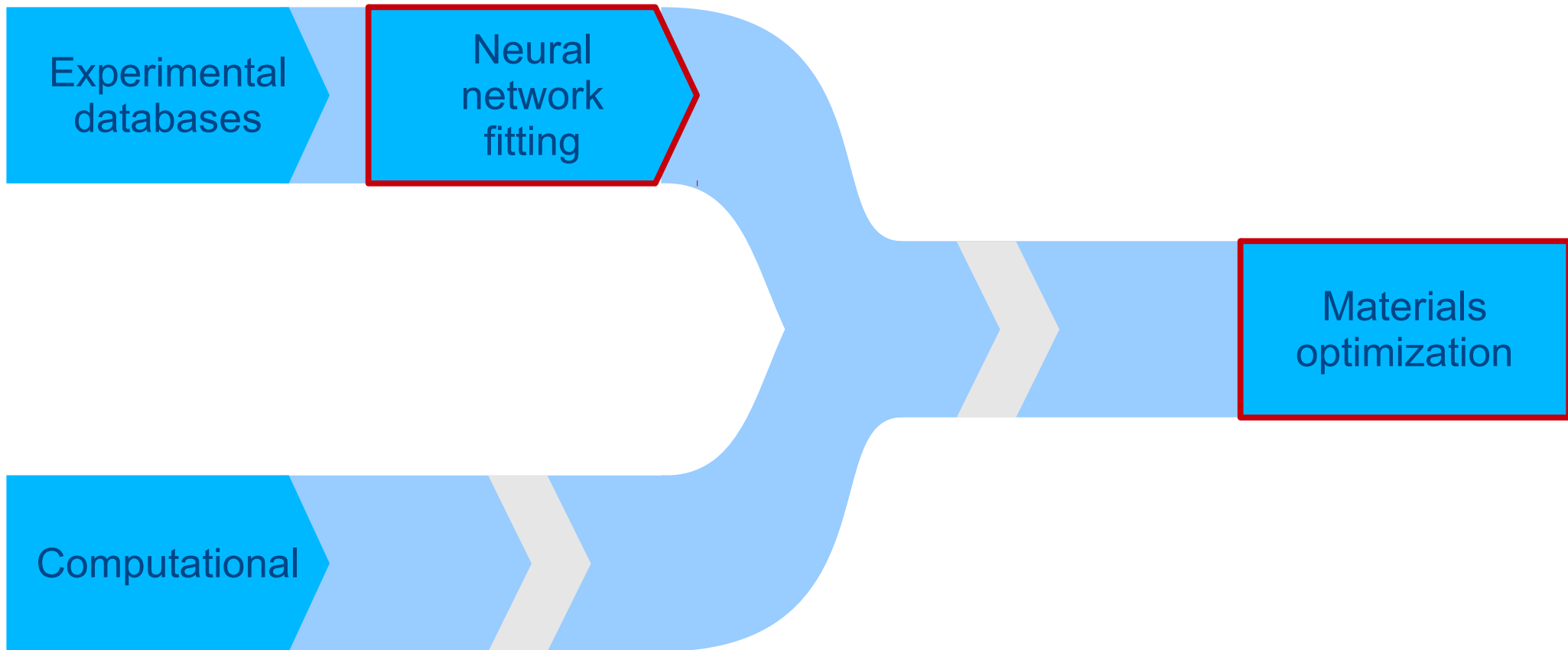
Probability



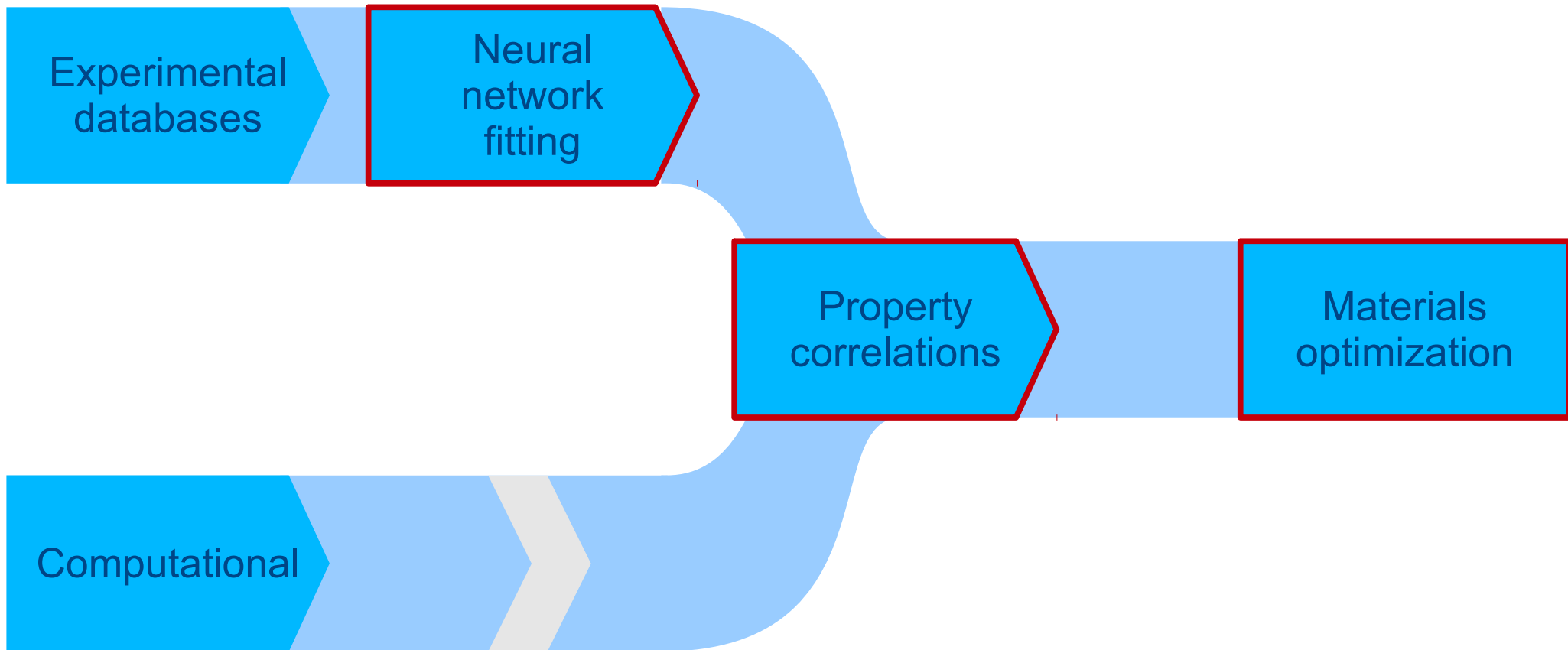
Ni-base superalloy



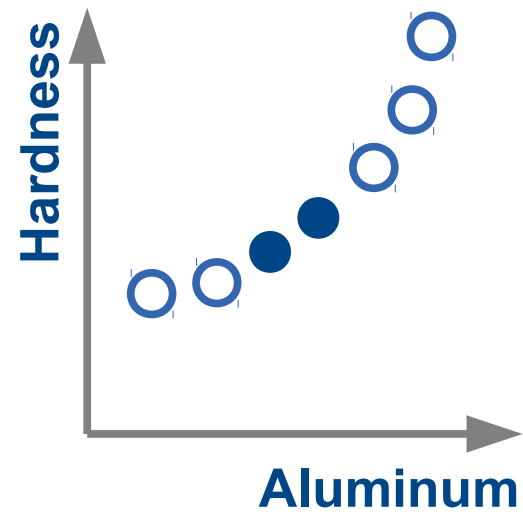
Two new tools



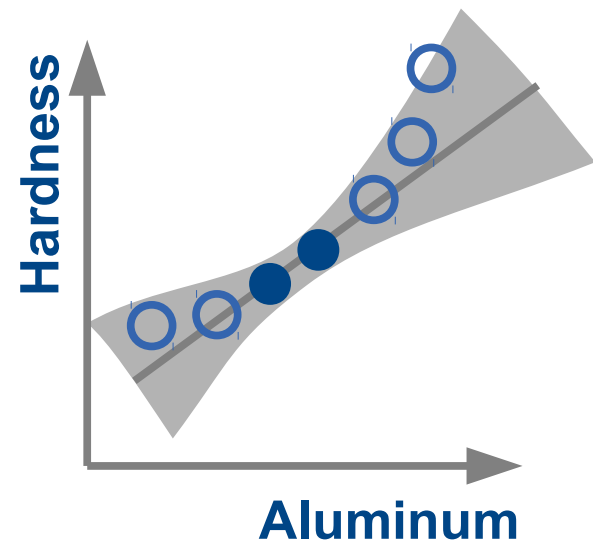
Three new tools



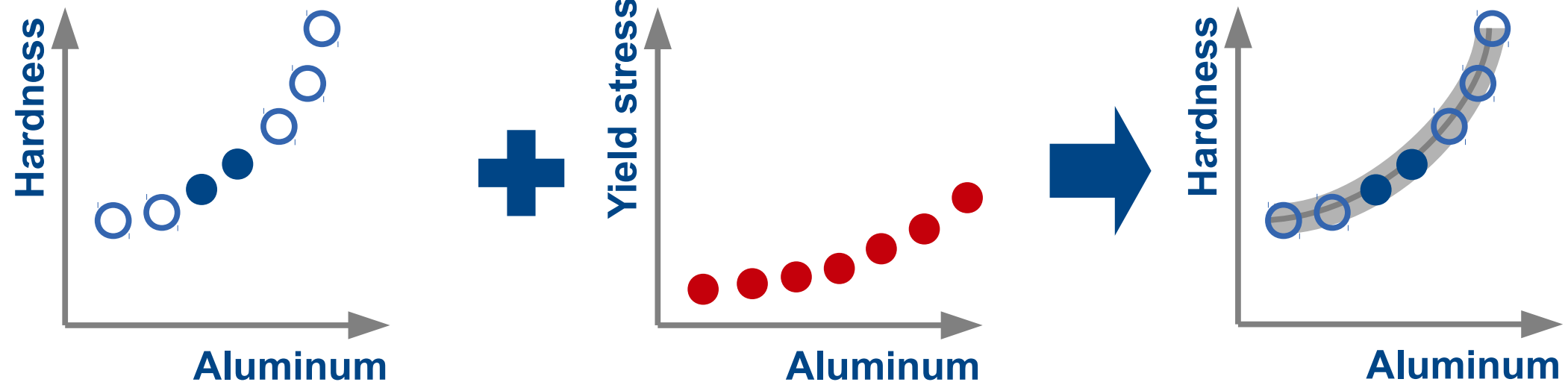
Correlations between properties



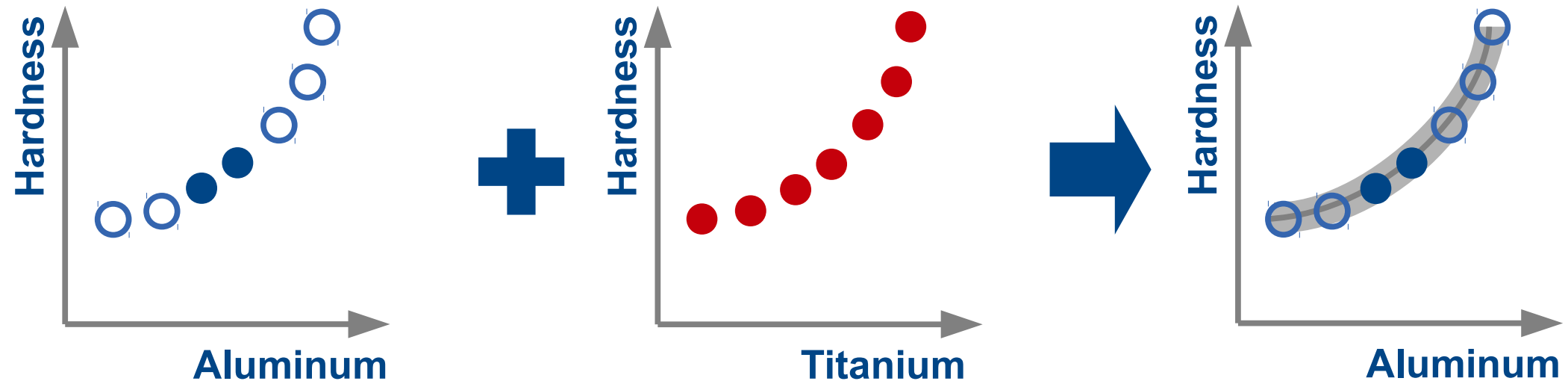
Correlations between properties



Correlations between properties



Correlations between properties

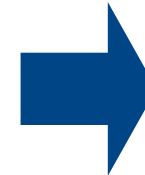


Exploiting correlations: 3D printing

**7 points for
3D printability**

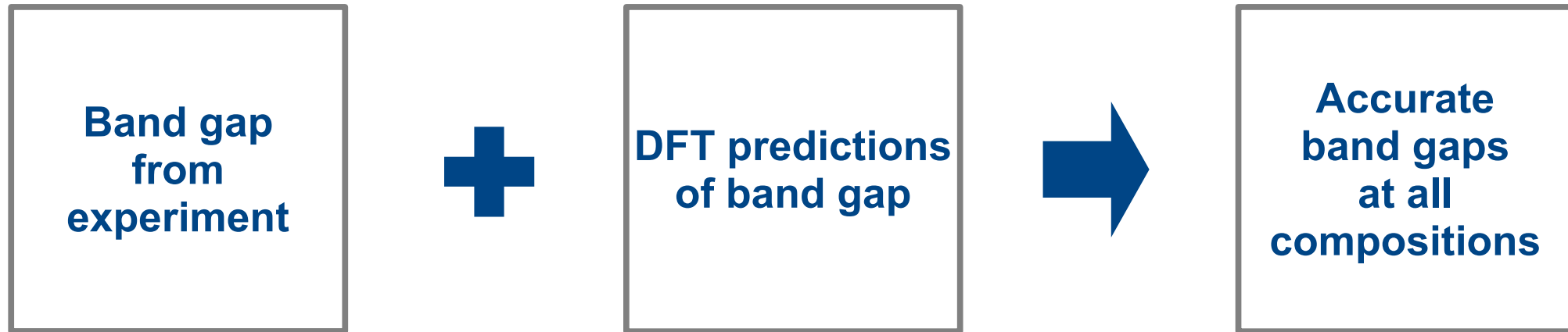


**Weldability
Heat capacity
Conductivity
Precipitates**

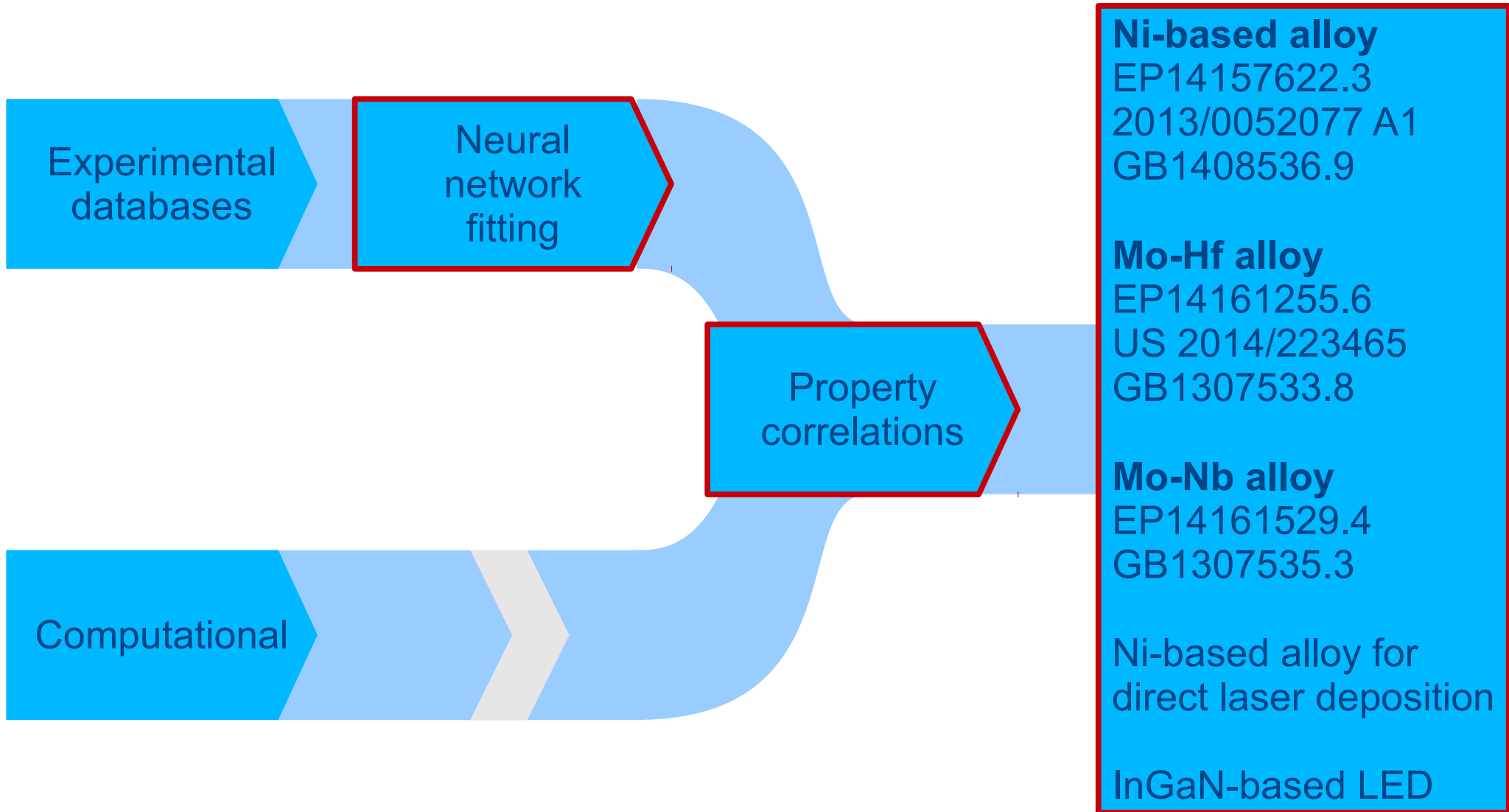


**Accurate
predictions for
3D printability**

Exploiting correlations: LEDs



Three new tools



Prospects in the future

Exploit correlations between material properties, compositions, and families to design four new alloys

Combine strengths of experimental databases with first principles approaches

Concurrent materials design

Recursive learning

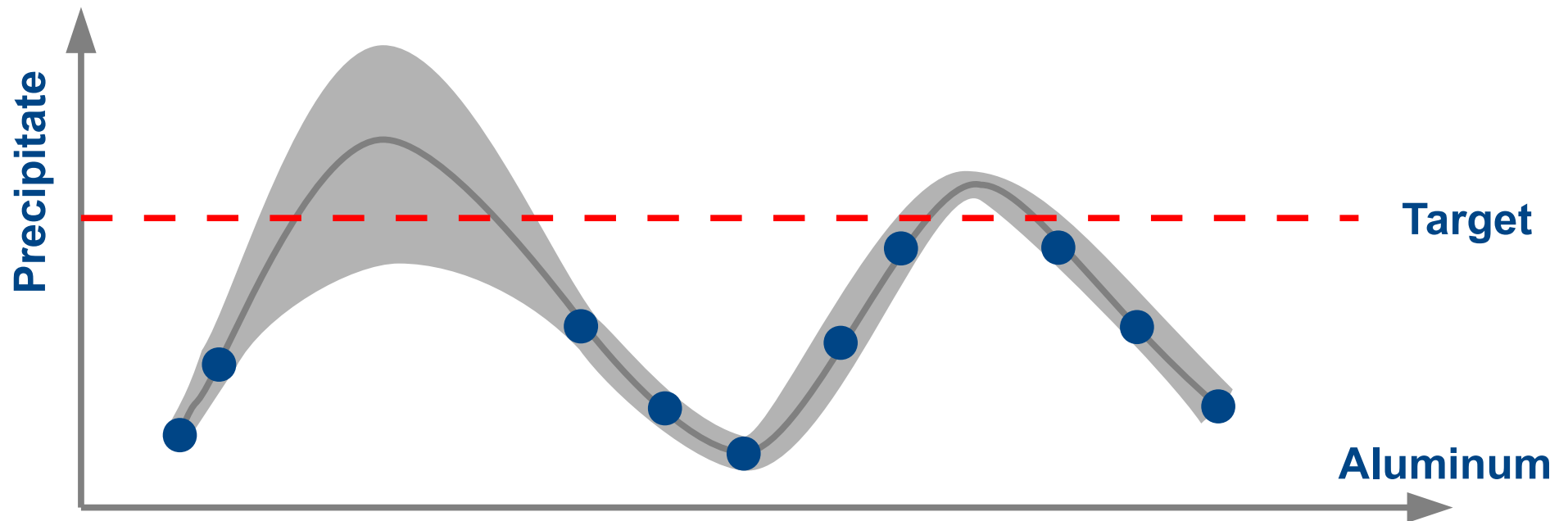
Calculate material property



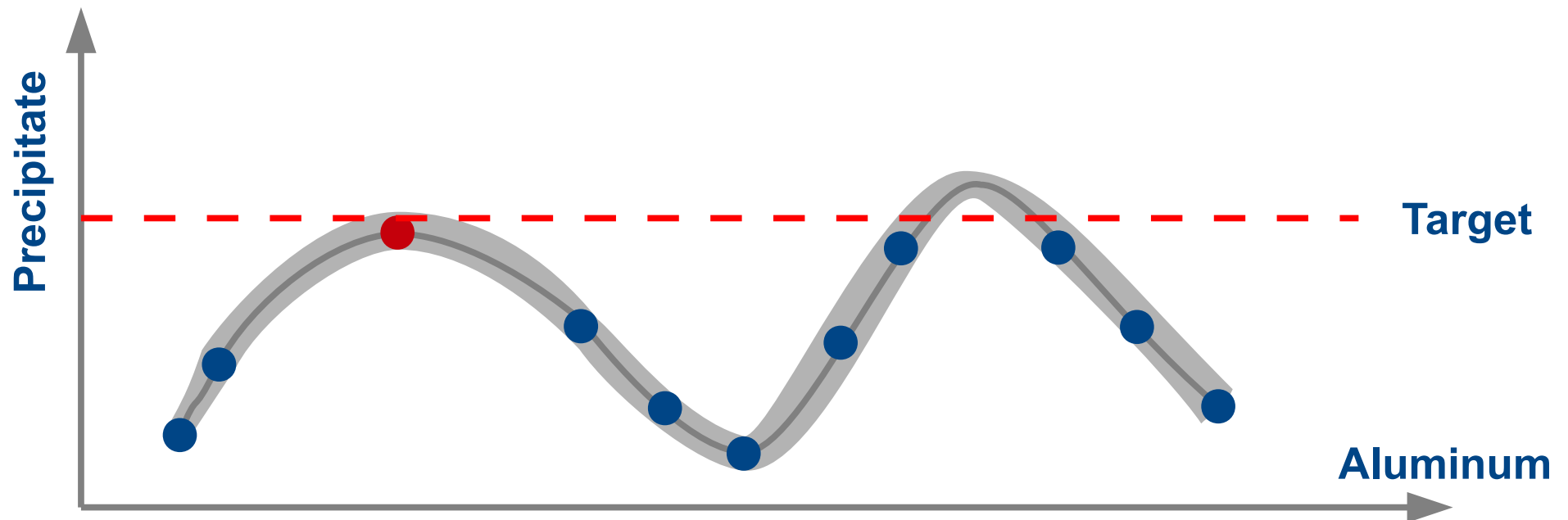
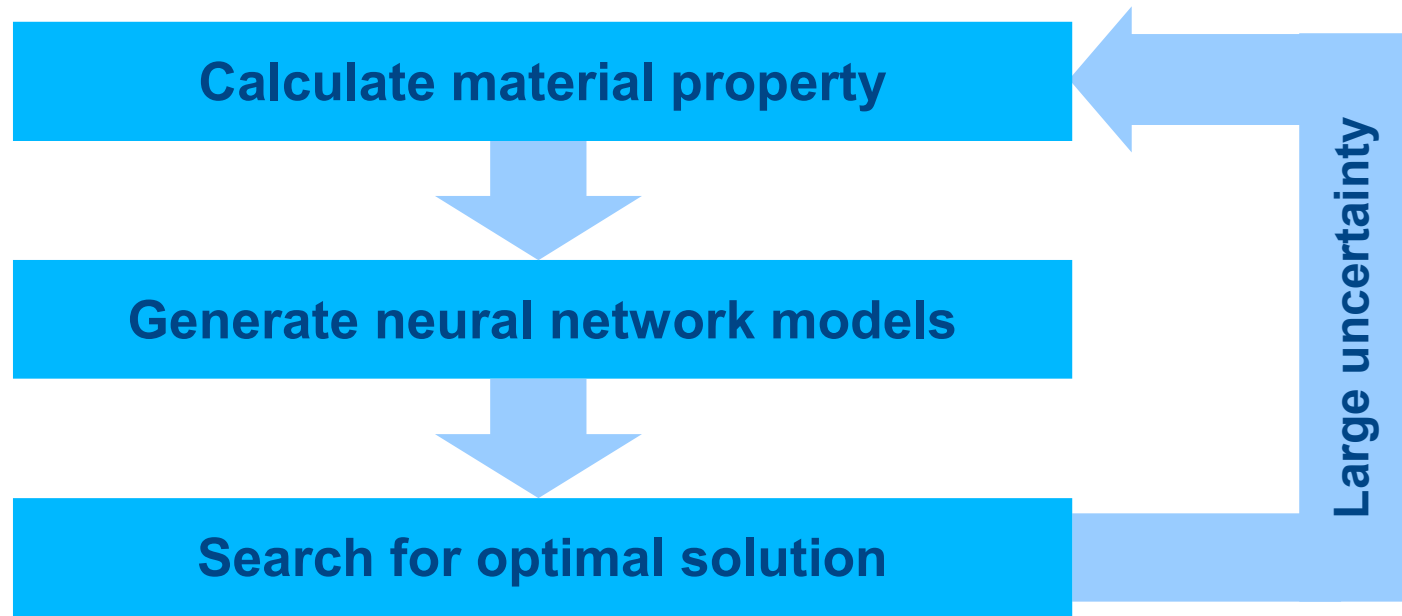
Generate neural network models



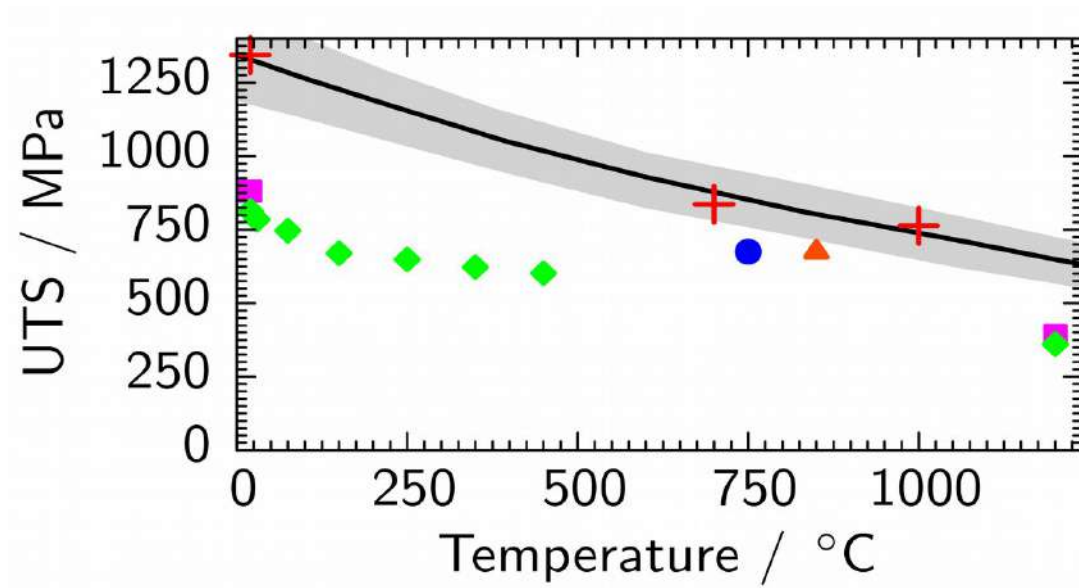
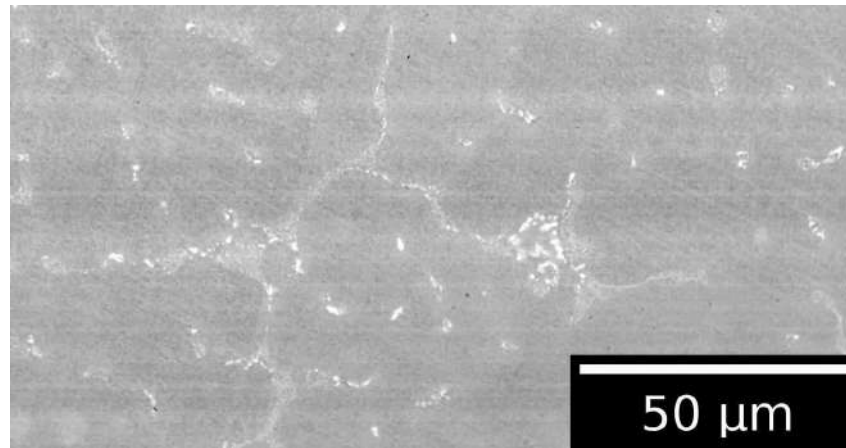
Search for optimal solution



Recursive learning



Mo-base alloy



- + Proposed
- MHC
- TZC
- ◆ TZM
- ▲ ZHM

Mo-base alloy

